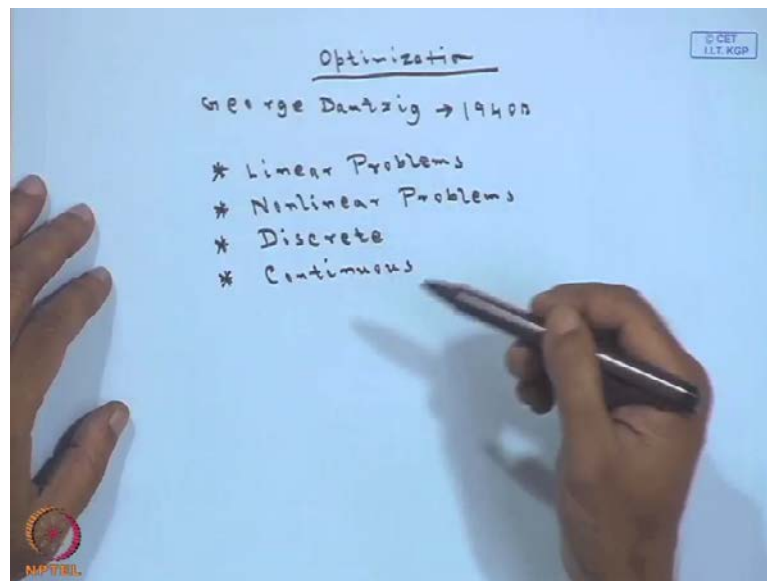


Optimization
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Lecture - 1
Optimization-Introduction

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In today's topics optimization this one, introduction to optimization on this one, today we are going to introduce some basic concepts of mathematical applications. Like matrix algebra, vector space, linear algebra which are required for us. Those tools are required for solving the problems on the optimization techniques. The first question, what comes to us is that, what is optimization? Basically, we say that optimization is a technique, is a mathematical application through which we can maximize or minimize a function. The function may be of one variable or may be of several variables. And when we are optimizing it; that is when we are maximizing or minimizing the function that may have subject to certain constraints.

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Machine Type	Product				Total time available per week
	P1	P2	P3	P4	
M1	2.7	3	4.6	3	3000
M2	2	7	2.5	1	9500
M3	2.4	4	6.1	3	6300
	7.5	4.6	9.2	5.25	

unit cost ← unit Profit

x_j be the number of units produced for product j

Minimize ← Maximize $Z = 7.5x_1 + 4.6x_2 + 9.2x_3 + 5.25x_4$

subject to

$$2.7x_1 + 3x_2 + 4.6x_3 + 3x_4 \leq 3000$$

$$2x_1 + 7x_2 + 2.5x_3 + 1x_4 \leq 9500$$

$$2.4x_1 + 4x_2 + 6.1x_3 + 3x_4 \leq 6300$$

Just like if you see, this particular table; if you see go through this one, in this table what we have done, some machines are there, if you see M 1, M 2, M 3 machines are there. These machines are producing some products P 1, P 2, P 3 and P 4. And per hour how much a production rate that we are telling that for machine 1 for producing P 1, what is the time required in terms of hours these are given. And total time available per week also is given. Unit profit for producing one product is also given, like for P 1 unit profit is 7.5. For P 2 it is 4.6 like this way. If you see, if we assume that x_j be the number of units produced for product j. In this case; of course, it is per week we are, I am talking about the per week, x_j be the number of units produced for product j. In that case; what actually I have to do? For product ne basically, I am producing x_1 quantity. For product 2 I am producing x_2 like this way. So, obviously; our aim is to maximize our profit always.

So, I can develop or formulate a mathematical model for this table like this way. Maximize Z equals $7.5x_1$ plus $4.6x_2$ plus $9.2x_3$ plus $5.25x_4$. So, this is actually our profit function. Because I am producing x_1 quantity of product 1 whose profit is 7.5. Similarly, for product 2, I am producing x_2 quantity, whose profit is 4.6 for unit, like this way. So, I want to maximize the value of this subject to certain constraints. What are the constraints? Subject to my constraint is; for each quantity of food item, how much time it takes? And how much total time available to hours or in other sense, you can tell $2.7x_1$ plus $3x_2$ plus $4.6x_3$ plus $3x_4$ this should be less than equals 3000. Similarly,

for machine 2; 2×1 plus 7×2 plus 2×3 plus 5.1×4 should be less than equals 9500. And 7.5×1 sorry, it should not be 7.5×1 , 2.4×1 plus 4×2 plus 6.1×3 plus 3×4 less than equals 6300.

So, if you see, our basic aim is; I want to find out how much quantity of each product should I produce per week? So, that my profit will be maxima. So, for that one I am formulating the model like this way. My profit function is this and subject to this constraints. So, once I am writing this basically, what is the job of optimization technique is that, how to find out the solution of this types of model. So, basically through optimization technique, what we will try to learn is that how to maximize or how to minimize a function subject to certain constraints. If you see instead of unit profit I have written here unit profit. If I write down this portion as unit cost as unit cost in that case instead of the maximization problem; obviously, I try to minimize my function and in that case your maximize will be reduced will be replaced by minimize.

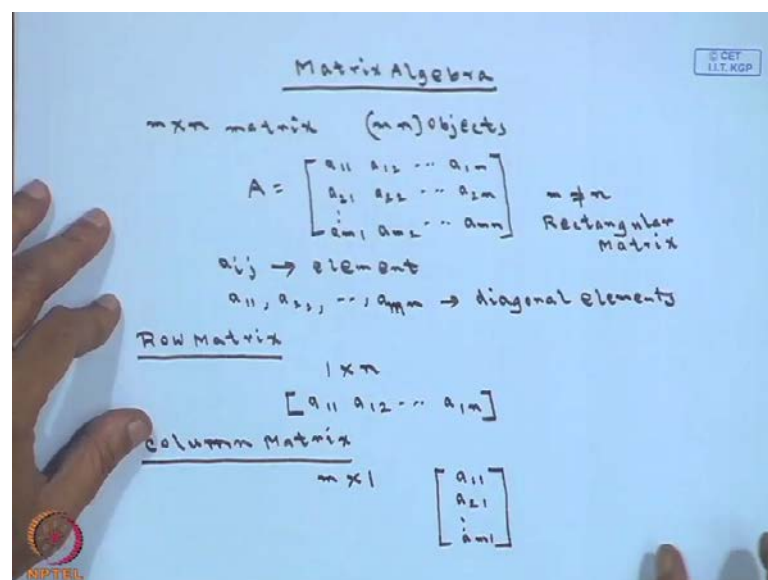
So, the problem may be maximization problem, the problem may be minimization problem. Depending up on what the data's are available. So, we want to learn the techniques through which we can solve this kind of problems. Basically, if you see George Dantzig, he developed this optimization; first optimization technique in 1940's to solve some problems related to military applications. That was the first thing he developed which we call as the linear programming problem. And that was the first optimization technique which was developed at the beginning. After that, lot of optimization models techniques had been developed for solving different types of problems. And these are widely used now a days in operational research, artificial intelligence, computer science and obviously, needless to say in industry it is being widely used. And which is helping a lot to the industry people.

Through these optimization techniques, we try to solve various kinds of problems. I can list the problems; one we call as the linear problems. Well, in the linear problems what we say, the optimization function basically should be linear. Whenever, the optimization or function or the constraints are linear functions then those types of problems we call it as the linear functions. Whereas, there is other type of problems, which I call non-linear problems, in non-linear problems whenever the optimization function or the constraints are non-linear in nature then those problems we call it as the non-linear problems.

We have the other type of problem that is we call it as discrete, in discrete case; what happens, the value of the variable will take some discrete integer values. Just like if you see whenever, I try to find out or made the class time table for teachers and I m locating subject to different rooms always they will takes some integer values. Or, in a hospital whenever, I am giving scheduling duties of the nurses those are taking the discrete values.

This kind of problems we call it as the discrete problems. There is another type which we call it as continuous problems, in continuous it can take any value any real value can be taken over there. And in this case; anything whatever you do, it can take 1.5, 1.10 something like this way, any real values it can take. So, we can categorize our problems linear, non-linear, discrete and continuous. And for each type of problem we have several different optimization techniques. We will go through one by one all these things. Of course, we will start with the linear problems initially and after that we will go to the non-linear and other type of problems also. Now, let us give some introduction as I told initially to the basic mathematical tools which will be required to us. You may know everything, but in brief we will tell the things. What is required or which we will be used afterwards in this lecture.

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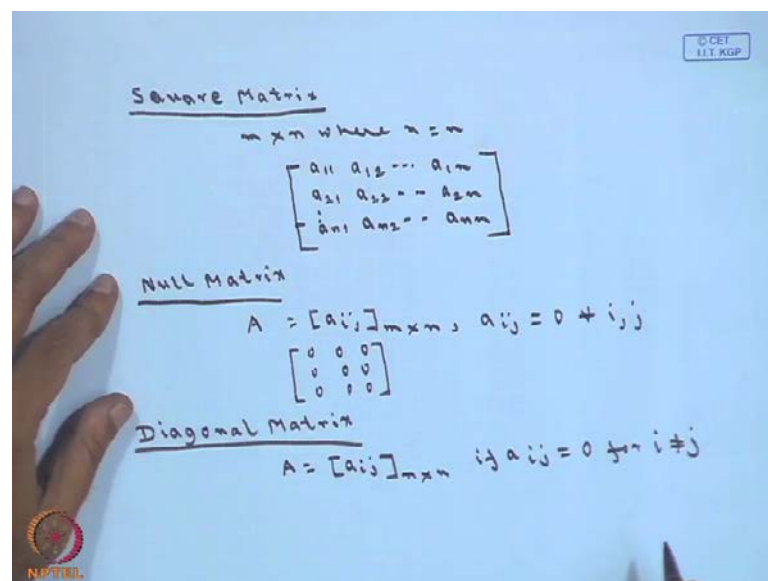


The first one is matrix algebra. In matrix algebra actually if you see, if you have m cross n matrix. This m cross n matrix is we say, is an arrangement of (m n) objects. This

objects may be distinct may not be distinct in and you are placing this $(m \times n)$ objects in m rows and n columns in a particular form. Then we call it as a matrix. So, whenever we are defining this m cross n matrix is basically nothing but; a arrangement of $(m \times n)$ objects in m rows and n columns and we write it in the form like this. $a_{11} a_{12} \dots a_{1n}$, $a_{21} a_{22} \dots a_{2n}$, $a_{m1} a_{m2} \dots a_{mn}$. So, we have total $(m \times n)$ objects. We say that the matrix is of order m cross n since, it has m rows and n columns. So, its order is m cross n . And whenever m is not equals n we tell that the matrix as rectangular matrix. Now, each object a_{ij} basically, these objects a_{ij} is we call it as an element of the matrix. That is the element on the i th row and on the j th column. This element we call it as the element of the matrix. Whereas, if you see $a_{11} a_{22}$ like this way a_{nn} that is this diagonals $a_{11} a_{22} \dots a_{nn}$ sorry, a_{mn} . This one we call it as the diagonal elements.

Now, Row matrix, just some definitions; a matrix of order 1 cross n , if I write down a matrix of order 1 cross n , that is 1 row and n columns. Then we call it as the row matrix. And we write it as $a_{11} a_{12} \dots a_{1n}$ like this way a_{1n} . So, this one is a row matrix. Similarly, you can define the column matrix; in case of column matrix what happens, where we have a matrix which takes the form of m cross 1 that is m rows and 1 column. Then we call it as the column matrix and it takes the form of this one; $a_{11} a_{21} \dots a_{m1}$ like this way it will go up to a_{m1} . So, $a_{11} a_{21} \dots a_{m1}$ this one we are calling as the column matrix.

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Next comes, square matrix; if you have a matrix of order m cross n , where m equals n . That is number of rows and numbers of columns are equal. Then we call that matrix as a square matrix of order n . When ever, you are having the number of rows and numbers of columns are same then we call that matrix as the square matrix. We can write down like this. So, this matrix is the square matrix which has n rows and n columns. Next here comes, null matrix; you have the matrix A equals a_{ij} , the order is m cross n . If a_{ij} equals 0 for all a_{ij} then that matrix we call it as the null matrix. I can write down something like this 0 0 0, 0 0 0 and all are 0. So, this matrix whatever I written is a null matrix of order 3 or it has 3 rows and 3 columns.

Next diagonal matrix, in a square matrix please note that, in a square matrix not rectangular matrix; that is square matrix means I am writing a_{ij} n cross n , if a_{ij} equals 0 for i not equals to j . That is, except the diagonal elements if all the elements are 0 then that matrix we call it as the diagonal matrix.

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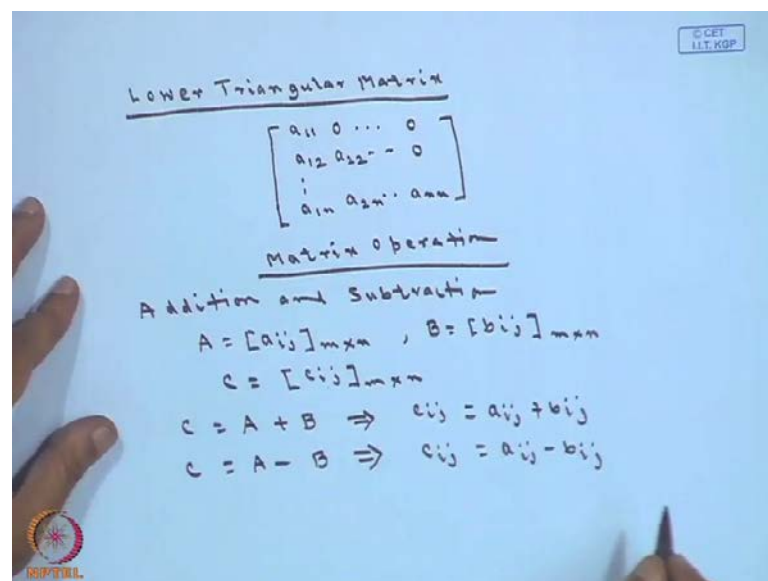
The image shows handwritten mathematical definitions on a blue background. At the top, a matrix is written as $\begin{bmatrix} a_{11} & 0 & \dots & 0 \\ 0 & a_{22} & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & a_{nn} \end{bmatrix}$. Below this, the text "Identity Matrix (unit Matrix)" is written. Underneath, the identity matrix is defined as $I = \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1 \end{bmatrix}$. Below that, the text "Upper Triangular Matrix" is written. At the bottom, a matrix is written as $\begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ 0 & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & a_{nn} \end{bmatrix}$. There are small logos in the bottom left and top right corners of the image.

Just like you can; for example, you can write down this matrix if you see, this one is the diagonal matrix except the diagonals all other elements are 0. So, this is a diagonal matrix. Next one is the identity matrix or sometimes we call it as the unit matrix also. In a diagonal matrix; if diagonal elements are 1 then that matrix we call it as the identity matrix. That is, these diagonal, values of these diagonal element if this is equals to 1 then we call it as the diagonal matrix. And this we may write it like this. This matrix where all

the elements are 0 except the diagonal elements. And this identity matrix we usually denote it by I which is the diagonal matrix or sometimes we call it as the identity matrix. The next one is upper triangular matrix; this upper triangular matrix is a square matrix where all the elements above and right of the diagonal matrix are non zero. Or, in other sense, you can tell all the elements below the diagonal elements are 0. In any way you can represent it either, all the elements above the diagonal elements and the right of the diagonal element are non 0.

So, for an example, you may say something like this. A matrix you consider this one. So, if you see this, in this case; all the elements above the diagonal elements are non zero and this side all the elements below the diagonal elements are 0. So, this one we call it as the upper triangular matrix.

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And, similarly you can define the lower triangular matrix also. The lower triangular matrix can be defined like this. I am directly giving one example, it will be clear to you I think. If you see this then all the elements above the diagonal elements are 0. And all the elements below these diagonal elements are non zeros. So, this kind of matrices we call it as the lower triangular matrices. So, basic matrix operations like addition subtraction. So, first come to the addition and subtraction. In addition and subtraction what happens, you first consider a matrix A a_{ij} , the size is of course, m cross n. There is another matrix B b_{ij} , whose size is also m cross n. I am considering the third matrix C, elements are c_{ij}

and this is also will be order this. Now, whenever you are writing C equals A plus B . It means one element $i j$ th element, c_{ij} of the resultant matrix this should be equals to a_{ij} plus b_{ij} . Similarly, whenever you are writing C equals A minus B , it implies c_{ij} this is equals to a_{ij} minus b_{ij} .

So, basically you are making term wise addition or term wise subtraction for performing the addition or subtraction operation. One thing should be noted that whenever, you are performing the addition or subtraction operation, the order of both the matrices should remain same it should not change over there.

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Matrix Multiplication

$A = [a_{ij}]_{m \times n}$, $B = [b_{ij}]_{n \times p}$, $C = [c_{ij}]_{m \times p}$

$C = [c_{ij}]_{m \times p} = AB = \begin{bmatrix} a_{i1} & a_{i2} & \dots & a_{in} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix} \begin{bmatrix} b_{11} & b_{12} & \dots & b_{1p} \\ b_{21} & b_{22} & \dots & b_{2p} \\ \vdots & \vdots & \ddots & \vdots \\ b_{n1} & b_{n2} & \dots & b_{np} \end{bmatrix}$

$c_{ij} = a_{i1} b_{1j} + a_{i2} b_{2j} + \dots + a_{in} b_{nj} = \sum_{k=1}^n a_{ik} b_{kj}$

Next comes matrix multiplication; in matrix multiplication what happens whenever, you try to multiply 2 matrices; the first thing required is that either the numbers of the columns of the first matrix should be equals to numbers of rows of the second matrix. Then only we can multiply 2 matrices otherwise we cannot multiply the matrices. So, again if you consider the matrices A equals a_{ij} say, whose order is m cross n . Let us take another matrix B which is equals to b_{ij} of order n cross p .

And, there is a third matrices C , C equals c_{ij} . If this is the resultant matrix; the order of this one will be m cross p . That is number of rows of the first matrix into number of columns of the second matrix. This will be the order of the resultant matrix. So, if you write down C equals this c_{ij} m cross p which is nothing but; again, here product of 2 matrices A and B . I am just writing this too. So, in details you have a_{11} a_{12} a_{1n} like

this way, you are having in middle i th row, i 1 i 2 i n and at last you are having a m 1 a m 2 a m n this is the matrix A . And you are having the matrix B similarly, b 1 1 here, I am writing b 1 j b 1 p , b 2 1 b 2 j b 2 p .

Similarly, b m 1 b sorry, this should not be n cross p . So, it will be n , b n j and b n p . So, whenever you are having 2 matrices like this. Sorry, just you see this one whenever your matrices, you have the matrices like this way; in that case what happens, you are c i j is nothing but; the element wise product and then some of these 2 rows and columns whatever I am marking. That is a i 1 into b 1 j plus a i 2 into b 2 j dot dot like this way plus a i n into b n j . I am writing, we can write down a i 1 b 1 j plus a i 2 b 2 j like this way, it will go and a i n into b n j . Or, in other sense in summation form k equals 1 to n a i k into b k j . So, this is the product of elements. So, like this way I can find out each element of the resultant matrix C , c i j using this formula.

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$$A = \begin{bmatrix} 4 & 0 \\ 6 & 0 \end{bmatrix}, B = \begin{bmatrix} 0 & 0 \\ 5 & 7 \end{bmatrix}$$

$$AB = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}, BA = \begin{bmatrix} 0 & 0 \\ 6 & 2 \end{bmatrix}$$

$$AB \neq BA \quad A \neq 0 \text{ or } B \neq 0$$

$$AB = AC \Rightarrow B = C$$

Transpose of a Matrix

$$A = [a_{ij}]_{m \times n} = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix}$$

$$A^T = \begin{bmatrix} a_{11} & a_{21} & \dots & a_{m1} \\ a_{12} & a_{22} & \dots & a_{m2} \\ \vdots & \vdots & \ddots & \vdots \\ a_{1n} & a_{2n} & \dots & a_{mn} \end{bmatrix}$$

$$(A^T)^T = A$$

$$(A+B)^T = A^T + B^T$$

$$(AB)^T = B^T A^T$$

If you take one example here, suppose you have the matrix A ; small matrix very 2 cross 2 matrix 4 0, 6 0. You have the matrix B 0 0 and 5 7. If I find out the product using our procedure then first term 4 into 0, 0 into 5 it plus 0, so it is 0. Like this way, you will find $A B$ is 0 0. Similarly, your $B A$ if you calculate, $B A$ also you can calculate because number of rows and number of columns of both the matrices are same. From your $B A$ you can calculate as, 0 0, 6 2 and 0. So, once I am calculating $A B$ and $B A$ one thing is very clear that is, it is not necessary that $A B$ will be always will be equals to $B A$. It may

be equals it may not be equals. Another thing is that; if product if you see here, although A and B 2 non zero matrices, the product of these 2 are 0.

So, if product of 2 matrices $A B$ equals 0, it does not necessarily imply that either A will be equals to 0 or B equals to 0. We should note this one. Similarly, if you are having $A B$ equals $A C$, it does not necessarily imply that B will be equals to C . It may hold it may not hold. So, these things we have to keep in mind. The next one is transpose of a matrix.

The transpose of a matrix is basically, whenever you are interchanging the rows and columns of a matrix and the resultant matrix we call it as the transpose of the matrix. That is, if you have a matrix A which you are telling as a $i \times j$ matrix, suppose you are having $a_{11} a_{12} a_{1n}$, $a_{21} a_{22} a_{2n}$ and $a_{m1} a_{m2} a_{mn}$, this matrix is there.

So, if I write down A transpose; then A transpose will be by simply by interchanging the number of rows and number of columns of this matrix. Means, here it will be $a_{11} a_{21}$ and then a_{m1} . This column is being changed as row. The next one will be $a_{12} a_{22}$ into a_{m2} and the last one will be $a_{1n} a_{2n}$ like this way a_{mn} . So, transpose of a matrix is a nothing but; whenever, you are interchanging the rows and columns of 2 matrices then we call it as the transpose of a matrix.

Transpose of a matrix has some properties like A transpose whole transpose if you take you will get back the original matrix itself. A plus B transpose this is equals to A transpose plus B transpose and $A B$ transpose this is equals the opposite way it will go, that is B transpose into A transpose. So, we will use these things for finding the solutions you will see how we are actually using all these properties afterwards. How you are representing, formulating a model converting in terms of matrices, you are trying find out the solutions. So, we are going to that part.

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The image shows handwritten notes on a blue background. At the top right, there is a small logo for 'CET T.T.KGP'. The notes are as follows:

Determinant
 $A = [a_{ij}]_{n \times n}$
 $\det(A) = |A| = \begin{vmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{vmatrix}$
 $|A| = \begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix} = a_{11}a_{22} - a_{12}a_{21}$

Minor
 $a_{ij} \quad |A|$

Cofactor
 $c_{ij} = (-1)^{i+j} M_{ij}$

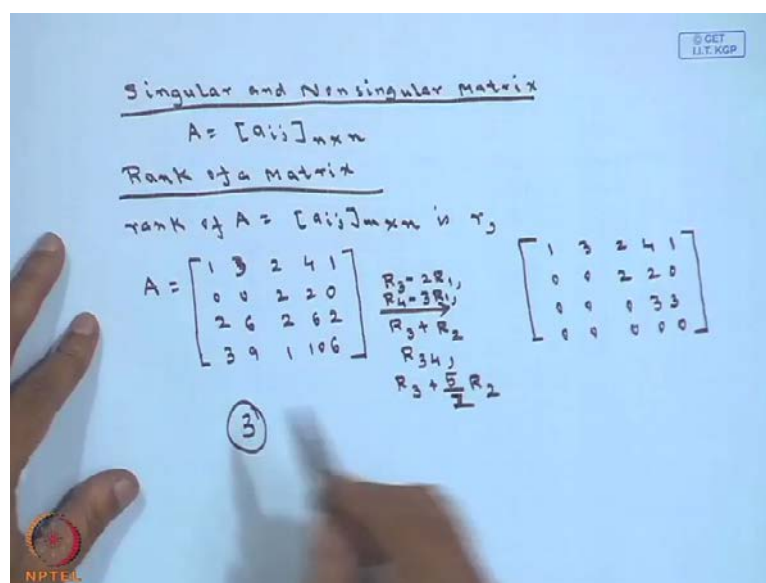
At the bottom left, there is a circular logo for 'NPTEL'.

The next one is determinant; consider a matrix A , A equals a_{ij} of order n cross n . That is it is a square matrix. Then $\det A$ we write denoted $\det A$ or this notion whatever, you want you can use. $\det A$ will be we write this way, a_{21} and lastly so $\det A$ this one. For example, if you have a 2 determinant of order 2 only that is $a_{11} a_{12}$, $a_{21} a_{22}$. I can evaluate it; I can obtain the value that is this element into this element minus this element into this element.

So, $a_{11} a_{22}$ minus a_{12} into a_{21} like this way you can evaluate 1 determinant and you can find a value. The next one is minor; the minor of an element a_{ij} obviously, of a determinate A , I am talking about with respect to A is the determinant of order n minus 1 by omitting the i th row and j th column. So, basically what is the minor? You have this determinant, if I want to find out the minor of a_{22} in that case; I will omit the second row and second column and I will get another determinant whose order will be 1 less.

Since, I have deleted 1 row and 1 column; order will be n minus 1. And that we call as the minor of the element a_{ij} . The next term is cofactor; the cofactor of an element again is defined as minus 1 to the power i plus j into M_{ij} . Where, M_{ij} is the minor of a_{ij} . So, once I know the minor, I can find out the cofactor of any element that is the minor into minus 1 to the power 1 plus i plus j .

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Next one is singular and non singular matrix; you have a matrix A which again, I am writing as n cross n matrix of square matrix of order n. Now, we say that A is a singular matrix if determinant of A is equals to 0. The matrix A is to be nonsingular matrix if determinant of A is not equals to 0. So, if determinant of A is not equals to 0 then the matrix is nonsingular, if it is equals to 0 then it is singular matrix. The next one is rank of a matrix; you have a matrix A obviously, A equals a i j of order m cross n.

We say that, the rank of A these is r. If all square minors of order r plus 1 or above are singular and there exist at least 1 minor of order r which is nonsingular. Then the rank of that matrix is r that is all the minors of order r plus 1 whose are singular. And at least 1 minor will be there order of r which is nonsingular or whose determinant does not vanish that we call as the rank of a matrix. Just let us take 1 example; you take a matrix A little bigger, this one for this matrix A.

Let us perform some row operations like this, 1 operation I will do R 3 minus 2 R 1 another 1 we will do R 4 minus 3 R 1. If I do these 2 operations R 3 minus 2 R 1 that is the first half first element of the third row will become 0. Similarly, the second element of the third row will be become 0 like this way you do the operations. If we go through this operations and after this I am not writing all of them because; it will take time I am just writing here, the operations you will perform 1 after another. That is first these 2 then you operate R 3 plus R 2. Then you interchange the third row fourth row and at last

you do this operations R 3 plus 5 by sorry, R 3 plus it should be 5 by 2 R 2 3 plus 5 by 2 R 2 these operations you do in order one after another in this order. And ultimately you will find 1 matrix of this particular form 1 3 2 4 1, 0 0 2 2 0, 0 0 0 3 3 and the last row has become 0.

So, if you see these; what is the rank of this matrix now? The original matrix was this, using row operations we transform this matrix into this one. You have the last row is this. So, if you take the rank of this matrix will be 3, the rank will become 3 because; if you take any matrix of order 4, that rank will be 0 over here because; 1 row is 0. So, if you evaluate the determinant value will be 0, but if you take 1 matrix of size is 1 determinant of order 3 which will be non zero. So, therefore; rank is 3. So, rank is the value for which the minor has non zero determinant or the minor is nonsingular this we call as the rank of the matrix.

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Adjoint of a Matrix

$A = [a_{ij}]_{n \times n}$

$\text{adj}(A) = \text{transpose of the matrix of cofactors of } A$

$A = \begin{bmatrix} 2 & -3 & 4 \\ 1 & 0 & 1 \\ 0 & -1 & 4 \end{bmatrix}$

cofactor of 2 = $(-1)^{1+1} \begin{vmatrix} 0 & 1 \\ -1 & 4 \end{vmatrix} = 1$

cofactor of -3 = $(-1)^{1+2} \begin{vmatrix} 0 & 1 \\ -1 & 4 \end{vmatrix} = -4$

$\text{adj}(A) = \begin{bmatrix} 1 & -4 & -1 \\ 8 & 8 & 2 \\ -3 & 2 & 3 \end{bmatrix}^T = \begin{bmatrix} 1 & 8 & -3 \\ -4 & 8 & 2 \\ -1 & 2 & 3 \end{bmatrix}$

The next one is Adjoint of a matrix; the Adjoint of a matrix, you have a matrix A, a i j be n square matrix of order n. Then Adjoint of A simply, we write as the transpose of just transpose of the matrix of cofactors of A. So, you are first finding out the cofactor of each element of the matrix A already, we have told how to find out the cofactor. And then make transpose of the matrix of the cofactors that matrix resultant matrix is called as Adjoint matrix. If you see, consider 1 example; A equals this, for this matrix cofactor of 2 just I will calculate 1 2 cofactor of the first element 2 is you delete the

corresponding row and column. So, it is minus 1 to the power $i + j$ that is 1 plus 1 into this matrix. If I delete this row and column, so it will be 0 1 and minus 1 4 if you evaluate it will become 1.

Similarly, you are cofactor of 3 also, cofactor of minus 3 this element this will be equals to minus 1 to the power 1 plus 2 into take the corresponding determinant after omitting the corresponding row and the corresponding column and you will get some value. Like this way I can find out the cofactors. So, Adjoint of a simply I can write down like this. You find out the cofactors 1 minus 4 minus 1 you can calculate of your own. And this one will come as this and transpose of this matrix that is you just write down this one as 1 8 minus 3 minus 4 8 2 and minus 1 2 and 3. So, this is the Adjoint of a matrix.

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Inverse of a Matrix

$A \rightarrow$

$$AB = BA = I$$

$$A^{-1} = \frac{\text{adj}(A)}{|A|}$$

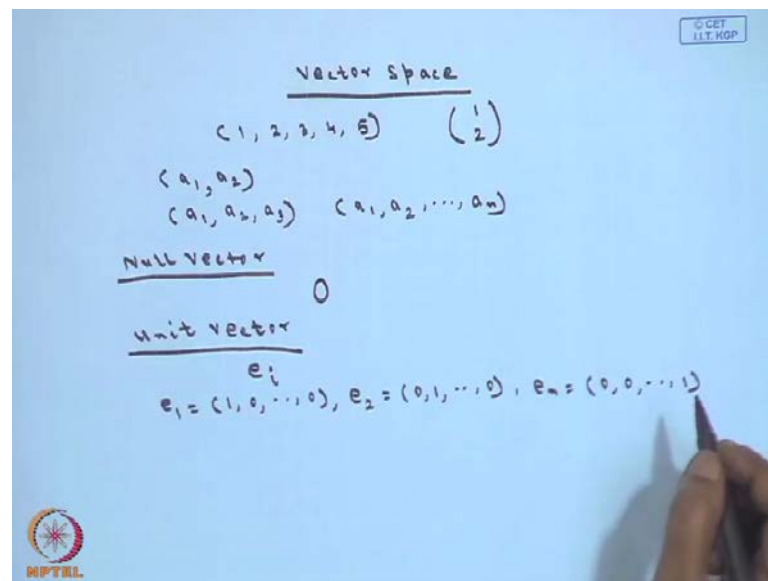
$$A^{-1} = \frac{1}{\det(A)} \text{adj}(A)$$

Next comes to the inverse of a matrix; the inverse of a matrix again, we are assuming that this matrix A is square matrix and it is a nonsingular matrix of order n . So, we say that matrix A is invertible, if there exist a nonsingular square matrix B their exist a nonsingular square matrix B . Such that B ; off course, the order of the B also will be n . Such that AB equals BA equals I . Then and we denote this A inverse by Adjoint of A by determinant of A .

So, actually what is happening; if the inverse of A matrix exist, if it is nonsingular, if it is square. Now, if we can find a matrix B of order n , square matrix B of order n . Such that, AB equals BA equals I . Where; off course, I is the identity matrix which we have told

earlier. Then A^{-1} we are defining as $\text{Adj}(A) / \det(A)$. For the earlier example whatever we told, A^{-1} can be calculated as this one, $1/10$ into $\text{Adj}(A)$ which already $\text{Adj}(A)$, we have calculated in the earlier example as this one. This is the $\text{Adj}(A)$ and the value of this determinant of this matrix A is 10. So, therefore; basically A^{-1} becomes this one. So, this is preliminaries about the matrix.

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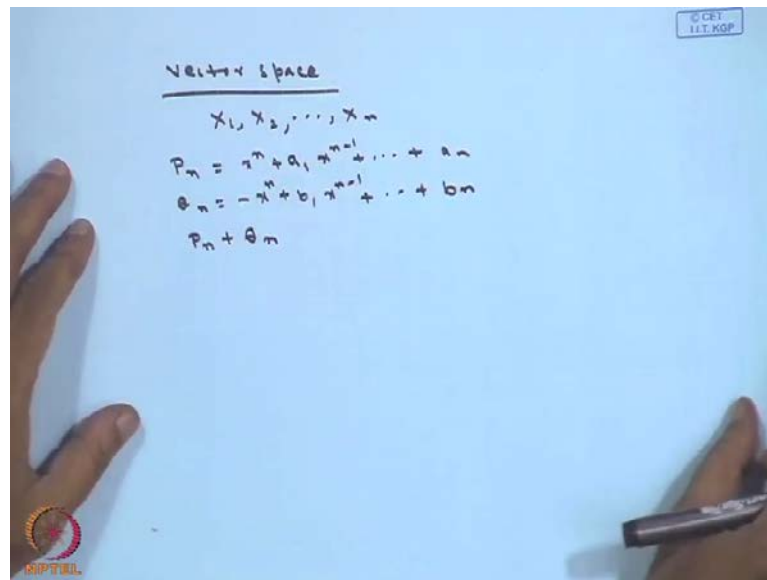


Now, let us come to the next one; that is vector space. Basically, if you say matrices having single row or column or often referred to as vectors; that is, if I write down something like this. This we call as a vector of size n equals 5 or number of components of this vector is 5. If you write something like this then; obviously, this is a column vector, this one was the row vector, this is the column vector with of size 2 since, it has 2 elements. If you see, the vectors whenever I am writing this a 1, a 2 this actually geometrically what I can say this represents a point in 2 dimensional space, when I am writing a 1 a 2 and a 3 this a point in 3 dimensional space. Similarly, if you are writing a 1, a 2 like this way a n then we can imagine it as a point on the n dimensional space.

Now, come to the null vector; null vector is a vector whose all the components and elements are 0 we denote it something like this way. So, null vector is the vector whose all components are 0. Next one is the unit vector; unit vector usually we denote it by the e_i is the e_i and what happens in the i th component the value will be 1 and all other component value will be 0 then we call it as the unit vector. So, when you are writing e

1; that means, first component will be 1, all other components will be 0. Similarly, if I am writing e_2 the second component will be 1 all the components will be 0. And if I am writing e_n then the last component will be 1 all other components will be 0 in this case. So, this is your unit vector.

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Now, come to the vector space; a set of vectors are there you assume a set of vectors x_1, x_2, \dots, x_n is there. This set of vectors x_1, x_2, \dots, x_n is closed under addition and scalar multiplication then; we say that this set of vectors form a vector space. So, if these vectors x_1, x_2, \dots, x_n are closed under addition and scalar multiplication then this forms a vector space. So, next question; obviously, comes what do you mean by closed under addition and scalar multiplication? A vector closed under addition means that, if you take any vector or 2 vectors of the set and if you add these 2 vectors the resultant vector should also be a member of the original set, Then we call it that it is your closed with respect to addition. Similarly, vector is closed or a set is closed with respect to scalar multiplication means, whenever you take a scalar multiply any vector the resultant vector should also be member of the original matrix, original vector. Then we say that it is form a vector space.

For example; if you try to take the examples, you take the set of all real numbers say; you take any 2 real numbers you form the addition, you will get the another number which is in the again a real number. Similarly, any 2 real numbers if I take and multiply I

will get another real number. So, set of real numbers are always forms a vector space. Similarly, if I take set of all polynomials whose degree is less than or equals to n will also form a vector space. But; consider the case set of all polynomials whose degree is exactly equals to n . Let us take a set of polynomial whose degree is equals to n ; that means, I can write down P_n equals something like this 1 polynomial $a_1 x$ to the power n minus 1 like this way a_n . And Q_n equals say, minus x to the power n plus 1 x to the power n minus 1 plus b_n .

Now, if you make the addition of these two polynomials P_n plus Q_n . In that case; you will find that, you are not you are getting 1 polynomial whose degree is n minus 1 but; my set was set of all polynomials of degree n . So, we can conclude that set of all polynomials will degree n will not form a vector space. So, like this way we can calculate the vector space and check whether a set is a vector space or not.

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Linear combination

$$a \in \mathbb{R}^n \quad a_1, a_2, \dots, a_n$$

$$a = \lambda_1 a_1 + \lambda_2 a_2 + \dots + \lambda_n a_n$$

$$\sum_{i=1}^n \lambda_i = 1 \quad \text{convex combination}$$

$$a_1 = (1, 2, 3), a_2 = (-1, 1, -1), a_3 = (0, 3, 2)$$

$$a_3 = 1 \cdot a_1 + 1 \cdot a_2$$

Linear dependence

$$a_1, a_2, \dots, a_n \in \mathbb{R}^n$$

$$\lambda_1 a_1 + \lambda_2 a_2 + \dots + \lambda_n a_n = 0$$

$$a_1 + a_2 - a_3 = 0$$

$$\lambda_1 = \lambda_2 = \dots = \lambda_n = 0$$

Linearly independent

$$a_1 = (1, 2)$$

$$a_2 = (-1, 1)$$

$$\lambda_1 a_1 + \lambda_2 a_2 = 0$$

$$\lambda_1 = \lambda_2 = 0$$

The next one is linear combination; a linear combination suppose, you have a vector a , which is in \mathbb{R}^n , if I write down a vector a is \mathbb{R}^n . It is a linear combination of vectors a_1, a_2, a_n . This a_1, a_2, a_n also belongs to \mathbb{R}^n , if I can write down a equals $\lambda_1 a_1$ plus $\lambda_2 a_2$ plus $\lambda_n a_n$ for some scalars λ . Sorry, for some λ_1, λ_2 and this will be λ_n for some scalars λ_1, λ_2 and λ_n . So, a vector a is there; that is another set of vectors a_1, a_2, a_n all of them belongs to \mathbb{R}^n . Then if I can write down a equal $\lambda_1 a_1$ plus $\lambda_2 a_2$ plus $\lambda_n a_n$ in this form. Where,

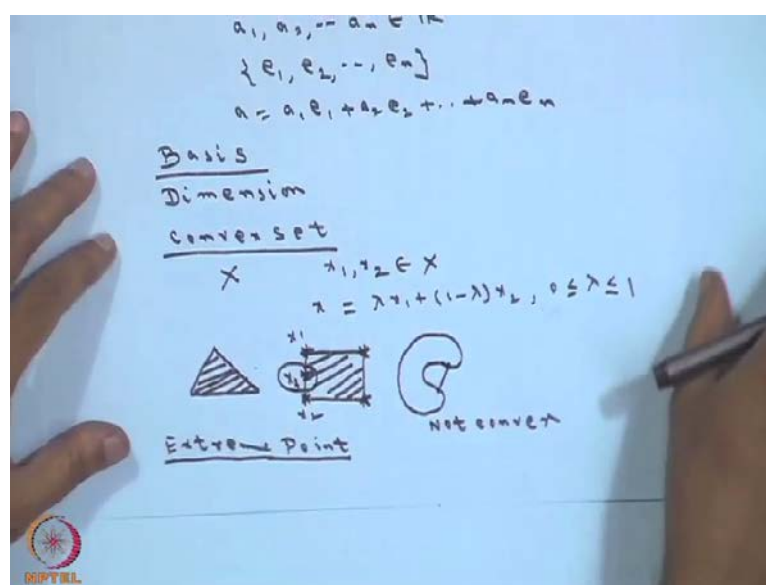
λ_i are our scalars then we say that \mathbf{a} is a linear combination of the vectors $\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n$. In particular, if $\sum_{i=1}^n \lambda_i = 1$ then this combination we call it as convex combination. Whenever, $\sum_{i=1}^n \lambda_i = 1$ this we call it as the convex combination.

For example, if you take a vector $\mathbf{a}_1 (1, 2, 3)$, \mathbf{a}_2 as $(-1, 1, -1)$ and \mathbf{a}_3 this equals to say $(0, 3, 2)$ always I can write down $\mathbf{a}_3 = 1 \mathbf{a}_1 + 1 \mathbf{a}_2$. So, your \mathbf{a}_3 sorry, this will be \mathbf{a}_1 . \mathbf{a}_3 will be a linear combination of \mathbf{a}_1 and \mathbf{a}_2 . Here, λ_1 is 1 and λ_2 also 1. Next comes linear dependence again, you had a set of vectors $\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n$ which belongs to, this set of vectors $\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n$ which is belongs to \mathbb{R}^n is said to be linearly dependent. If it said to be linearly dependent, if there exist scalars λ_i . Such that, $\lambda_1 \mathbf{a}_1 + \lambda_2 \mathbf{a}_2 + \dots + \lambda_n \mathbf{a}_n = \mathbf{0}$. But not all λ_i is 0.

Then, we say the vectors are linearly dependent. So, a set of vectors $\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n$ are linearly dependent. Whenever, $\lambda_1 \mathbf{a}_1 + \lambda_2 \mathbf{a}_2 + \dots + \lambda_n \mathbf{a}_n = \mathbf{0}$ where, all λ_i are not equals to 0. So, for the earlier example, this example $\mathbf{a}_1, \mathbf{a}_2, \mathbf{a}_3$ these are linearly independent. Sorry, linearly dependent. Since, $\mathbf{a}_1 + \mathbf{a}_2 - \mathbf{a}_3 = \mathbf{0}$ and all λ_i are not equals 0. But on the other way, if $\lambda_1 \mathbf{a}_1 + \lambda_2 \mathbf{a}_2 + \lambda_n \mathbf{a}_n = \mathbf{0}$ this is equals 0. This holds when all λ_i is 0. That is $\lambda_1 = \lambda_2 = \dots = \lambda_n = 0$. Then we say that the vectors $\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n$ are they are linearly independent. I am just writing this, they are linearly independent.

So, these two concepts are very independent. one is linearly dependent and another one is linearly independent. So, linearly independent means, whenever you are make the linear combination of this that is equals to 0, but all the λ_i are equals to 0. If you consider; if I am just writing here, one vector \mathbf{a}_1 and another vector equals this. So, if I take $\lambda_1 \mathbf{a}_1 + \lambda_2 \mathbf{a}_2 = \mathbf{0}$ this is equals 0. And this is possible only when $\lambda_1 = \lambda_2 = 0$. So, I can tell that the vectors \mathbf{a}_1 sorry, the vectors \mathbf{a}_1 and \mathbf{a}_2 are linearly independent.

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Next there comes spanning set; again take a set of vectors a_1, a_2, \dots, a_n which belongs to \mathbb{R}^n . A set of vectors $a_1, a_2, \dots, a_n \in \mathbb{R}^n$ is said to span or generate \mathbb{R}^n if every vector of \mathbb{R}^n can be written as a linear combination of a_1, a_2, \dots, a_n . That is you are having a set of vectors from this set of vectors you are generating all the vectors means, any vector you take in \mathbb{R}^n that will be a linear combination of the vectors a_1, a_2, \dots, a_n . Then this set through which you are generating we call it as the spanning set. And obviously; this set of vectors are linearly independent vectors, this a_1, a_2, \dots, a_n this spanning set should be linearly independent. For example, if you want to take the example in that case; if you take the set of vectors e_1, e_2, \dots, e_n . If you take these vectors any vector on \mathbb{R}^n can be generated from this set e_1, e_2, \dots, e_n . So, your e_n is $1, 0, 0$ unit vectors on unity it is 1 on the first component, all others are zero's.

So, any vector a can be represented in terms of this one that is you can write down $a = a_1 e_1 + a_2 e_2 + \dots + a_n e_n$. So, this is the spanning set of this. Next there comes basis of a vector; a basis we call it linearly independent set of vectors which spans the \mathbb{R}^n , which spans the vector space. In our earlier examples; e_1, e_2, \dots, e_n these are the basis of the \mathbb{R}^n . So, basically the linearly independent set of vectors which spans all the elements all the vectors of set is called the basis of that. For this case; e_1, e_2, \dots, e_n is the basis of this case. And along with this there is another one is which we call as the dimension of a set. Dimension means, number of linearly independent vectors whatever is there that number we call it as the basis. Now, next comes to convex set; you

have a set X , the set X as we call as convex set. If for any points say x_1, x_2 which belongs to X , if you join the line segment of the x_1 to x_2 the line segment is also lies on X then we say that the set is a convex set.

So, you have any 2 points; x_1, x_2 of X . If you join the line segment x_1, x_2 , if inter line segment lies on the set X then X is called the convex set. Mathematically, if I have to say then we will say that, if x_1, x_2 belongs to X . If I can find a point a, x that x equals $\lambda x_1 + (1 - \lambda)x_2$ where λ lies between 0 to 1. If it is satisfied then the set is a convex set. For example, if you consider; if you take a triangle that triangle and its interior will form 1 convex set because if you take any point, any 2 points inside these if you join that will lie also inside the triangle.

Similarly, if I take a point something like this, in that case; also this will form a convex set this entire rectangle place. But suppose, I have taken something like this, some figure like this, this is not a convex set because; if I take a point here, if I take a point here, if I join these line segment does not lie on this set. Therefore, this one is not a convex set.

Similarly, there is just another point that is extreme point. And extreme point of a convex set X . This extreme point of a convex set will be that point such that, we cannot find 2 other points x_1 and x_2 . Such that, if I join them I will get the point or in other sense, if I have to say this point if you consider this point, this point these are the extreme points, because I cannot find out any other 2 points which will lie over here. But if you consider this x_1 say x_2 if you take another point x_3 although it is on the exterior. But I am finding 2 points x_1, x_2 and if I joining them then x_3 lie on X . Therefore, x_3 is not a extreme point whereas, x_1 and x_2 is the extreme point.

Thank you.