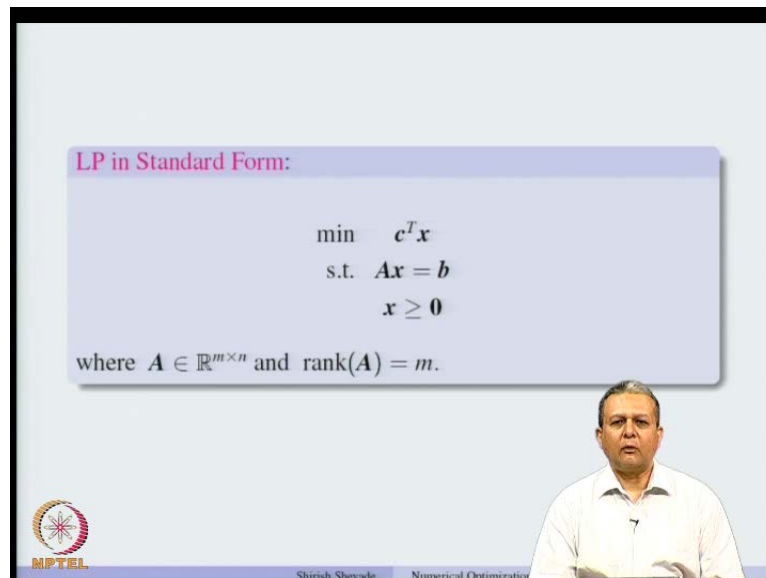


Numerical Optimization
Prof. Shirish K. Shevade
Department of Computer Science and Automation
Indian Institute of Science, Bangalore

Lecture - 33
Optimality Conditions and Simplex Tableau

Hello, welcome back to this series of lectures on Numerical Optimization. In the last class we saw the equivalence between an extreme point of a feasible set, and the vertex basic feasible solution of a linear program. Now, in today's class we will discuss about the solution or an algorithm to get a solution of a linear program.

(Refer Slide Time: 00:51)



LP in Standard Form:

$$\begin{aligned} \min \quad & c^T x \\ \text{s.t.} \quad & Ax = b \\ & x \geq 0 \end{aligned}$$

where $A \in \mathbb{R}^{m \times n}$ and $\text{rank}(A) = m$.

NPTEL
Shirish Shevade Numerical Optimization

So, let us quickly relook at some of the topics that we discussed in the last class. So, we started with the linear programming standard form, which is of the type minimize c transpose x subject to x equal to b and x non negative, and A is rank m matrix of size m by n .

(Refer Slide Time: 01:14)

Let x be a *nondegenerate basic feasible solution* corresponding to the basic variable set B and non-basic variable set N .
 Let B denote the basis matrix.

$$Ax = b$$


$$\therefore Bx_B + Nx_N = b$$

$$\therefore x_B = B^{-1}b - B^{-1}Nx_N$$

Particular Solution: $x_B = B^{-1}b$ and $x_N = 0$

Objective Function = $c^T x$
 $= c_B^T x_B + c_N^T x_N$
 $= c_B^T B^{-1}b - c_B^T B^{-1}Nx_N + c_N^T x_N$
 $= \bar{z} + \bar{c}_B^T x_B + \bar{c}_N^T x_N$

where $\bar{c}_B^T = 0^T$ and $\bar{c}_N^T = c_N^T - c_B^T B^{-1}N$ are the *relative cost factors* corresponding to the basis matrix B and \bar{z} denotes the current objective function value.



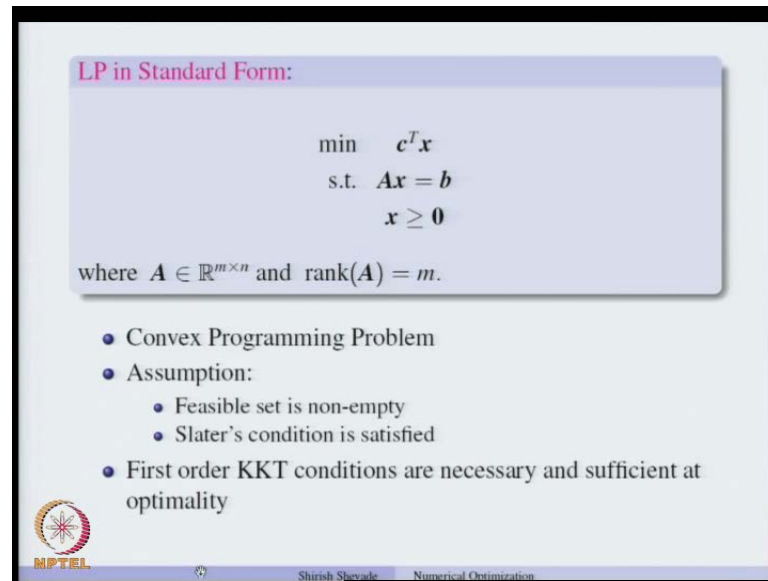
Shrish Shekhar Numerical Optimization

And we assume that x is a non degenerate basic feasible solution, which corresponds to the basic variable set B , and the non basic variable set N . And if B denotes the basis matrix, then $Ax = b$ can be written in the form $Bx_B + Nx_N = b$. So, A is split in to two sub matrices B and N , and the corresponding components of x are x_B and x_N . And they put together will give us this equation $Bx_B + Nx_N = b$, and the since b denotes the basis matrix that means, the columns of b are linearly independent.

And is of the size m by N , we can always invert it and get a solution x_B which is of this form, $x_B = B^{-1}b - B^{-1}Nx_N$, and a particular solution is piloting x_N to be 0 in this solution, we get $x_B = B^{-1}b$ and $x_N = 0$. And the objective function $c^T x$ can be written as the current objective function, which is $c_B^T B^{-1}b$, which we are going denote by \bar{z} plus $c_B^T x_B$ plus $c_N^T x_N$.

Now, \bar{c}_B is 0 and \bar{c}_N is nothing but $c_N^T - c_B^T B^{-1}N$, these are called the relative cost factors, associated with basis matrix B . And the current objective function value is $c_B^T B^{-1}b$, because \bar{c}_B is 0 and x_N give 0. So, the current objective function is nothing but \bar{z} which is $c_B^T B^{-1}b$.

(Refer Slide Time: 02:52)




LP in Standard Form:

$$\begin{aligned} \min \quad & c^T x \\ \text{s.t.} \quad & Ax = b \\ & x \geq \mathbf{0} \end{aligned}$$

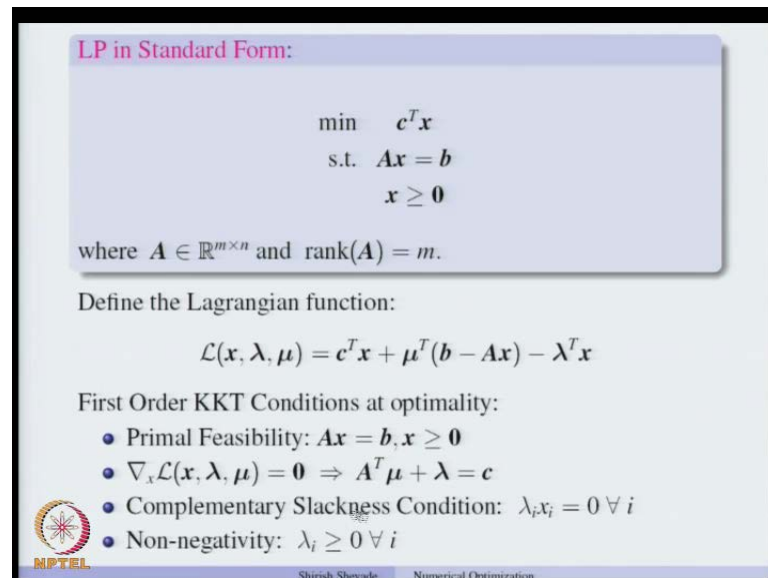
where $A \in \mathbb{R}^{m \times n}$ and $\text{rank}(A) = m$.

- Convex Programming Problem
- Assumption:
 - Feasible set is non-empty
 - Slater's condition is satisfied
- First order KKT conditions are necessary and sufficient at optimality

 Shirish Shevade Numerical Optimization

So, we also assume that the feasible set of this linear program is non empty, and also Slater's condition is satisfied that means, the feasible set has non empty interior. Then under this condition, we had already seen that first order KKT conditions, are necessary and sufficient at optimality.

(Refer Slide Time: 03:17)



LP in Standard Form:

$$\begin{aligned} \min \quad & c^T x \\ \text{s.t.} \quad & Ax = b \\ & x \geq \mathbf{0} \end{aligned}$$


where $A \in \mathbb{R}^{m \times n}$ and $\text{rank}(A) = m$.

Define the Lagrangian function:

$$\mathcal{L}(x, \lambda, \mu) = c^T x + \mu^T (b - Ax) - \lambda^T x$$

First Order KKT Conditions at optimality:

- Primal Feasibility: $Ax = b, x \geq \mathbf{0}$
- $\nabla_x \mathcal{L}(x, \lambda, \mu) = \mathbf{0} \Rightarrow A^T \mu + \lambda = c$
- Complementary Slackness Condition: $\lambda_i x_i = 0 \forall i$
- Non-negativity: $\lambda_i \geq 0 \forall i$

 Shirish Shevade Numerical Optimization

So, we write down the Lagrangian of this objective this optimization problem, which is nothing but the L of x lambda mu. Now, lambda's are the Lagrangian multipliers associated with the inequality constraints, and mu's are the Lagrangian multipliers

corresponding to the equality constraints. So, the Lagrangian function is the objective function $c^T x$ plus $\mu^T b$ minus $\lambda^T (Ax - b)$. Now, if you write down the KKT conditions, we will see that, the first order KKT conditions at optimality or that the primal x is primal feasible.

So, which means it should satisfy x equal to b and x non negative, then the gradient of the Lagrangian with respect to x should vanish at optimal at optimal point x, λ, μ . So, with that implies that $c - A^T \mu - \lambda = 0$, so which gives us $A^T \mu + \lambda = c$, and complimentary slackness condition $\lambda_i x_i = 0$ for all i . Now, there are also a non negativity constraint that is $\lambda \geq 0$. So, all these conditions should be satisfied by any optimal point x, λ, μ . And in today's class, we will see how to get a solution of the linear program by making use of this optimality conditions.

(Refer Slide Time: 04:58)

Let x be a *nondegenerate basic feasible solution* corresponding to the basic variable set B and non-basic variable set N .
 $x = (x_B \ x_N)^T$ where $x_B > 0$ and $x_N = 0$.
 At optimal (x, λ, μ) ,


- $\lambda_B = 0$ and $\lambda_N \geq 0$.
- $c = A^T \mu + \lambda$. That is,

$$\begin{pmatrix} c_B \\ c_N \end{pmatrix} = \begin{pmatrix} B^T \\ N^T \end{pmatrix} \mu + \begin{pmatrix} \lambda_B \\ \lambda_N \end{pmatrix} \Rightarrow \begin{matrix} c_B = B^T \mu + \lambda_B \\ c_N = N^T \mu + \lambda_N \end{matrix}$$

- $\lambda_B = 0 \Rightarrow c_B = B^T \mu \Rightarrow \mu = B^{-T} c_B$
- $\lambda_N \geq 0$ requires that

$$\lambda_N = c_N - (B^{-1}N)^T c_B \geq 0$$

The current basic feasible solution x is *not* optimal if there exists $x_q \in N$ such that $\lambda_q < 0$.



Shrish Shrivastava | Numerical Optimization

Now, if x is non degenerate basic feasible solution, so let us write x as x_B and x_N , as two parts of x , where x_B is strictly greater than 0, and x_N equal to 0. So, as we saw at optimal x, λ, μ , since at any feasible x , basic feasible x , we are assuming that there is non-degenerate that means, that x_B is greater than 0. So, if that particular x is optimal, then what we want is that the λ_B should be 0, because of the complimentary slackness condition; and since, x_N is 0, λ_N can be greater than or equal to 0.

So, if we are given x , we will first find out the value of μ and see whether λ satisfies the KKT conditions, so if they do satisfy KKT conditions, then we can claim that we have found the solution. Now, remember that μ 's are the Lagrangian multipliers associated with equality constraints and therefore, they are unrestricted in sign. On the other hand, λ are the Lagrangian multipliers associated with the inequality constraints of the type x greater than or equal to 0.

And therefore, λ are restricted in sign, and we make use of that too check whether a given point x λ μ is a optimal point. And therefore, we started with a basic feasible point or basic feasible solution which is x_B x_N , which if it is optimal λ_B has to be 0, and λ_N non negative. Now, c is equal to $A^T \mu + \lambda$, now we will make use of this equation to get μ or in other words, if we as we split x into two parts x_B and x_N , we split c also into two parts c_B and c_N .

And the A matrix we have already split it into two parts B and N , so that the equation $c = A^T \mu + \lambda$, can be written as $c_B = b^T \mu + \lambda_B$, and $c_N = N^T \mu + \lambda_N$. Now, now if λ_B is equal to 0 that means, that we can write c_B is equal to $b^T \mu$ and μ is equal to $b^T \mu$ inverse c_B , so that gives the value of μ . So, for a given x we have found out μ , we have also we also want λ_B to be 0 so that, KKT conditions would be satisfied.

And the m is to check whether λ_N is are 0, so by forcing λ_B to be 0, if we can get λ_N to be greater than or equal to 0, then the point x λ μ would satisfy the optimality conditions. So, by forcing λ_B to be 0, we have got μ and now the next step is to get λ_N , now λ_N is nothing but $c_N - N^T \mu$, and if you plug in this value of μ what we get is that, λ_N is equal to $c_N - N^T c_B$. And that should be greater than or equal to 0, so which means that the the Lagrangian multipliers corresponding to the non basic variables N , they should be non negative.

So, if we look at of the steps that we followed, so we started with the basic feasible solution x which consisted of x_B and x_N , where x_B is greater than 0, and x_N equal to 0. Now, in order to check this solution is optimal, what we need to is that since x_B is greater than 0, we can force λ_B to be 0, because complimentary slackness

condition needs to be satisfied. And x_N equal to 0 and therefore, λ_N had to be non negative and by using the fact that λ_B equal to 0 we got the value of μ , and we use that value of μ to be find λ_N .

And now it is just a matter of checking whether λ_N is greater than or equal to 0 for all non negative variables or non basic variables. So, the current basic feasible solution x is not optimal, if there exists a non basic variable say x_q such that, the corresponding Lagrangian multiplier is less than 0 or strictly negative. So, in such a case we can definitely say that, the current solution current basic feasible solution is not optimal.

(Refer Slide Time: 10:13)

x is feasible $\Rightarrow Ax = b$.

$\therefore Bx_B + Nx_N = b \Rightarrow x_B = B^{-1}b - B^{-1}Nx_N$

Objective Function at $x = c^T x$

$$= c_B^T x_B + c_N^T x_N$$

$$= c_B^T (B^{-1}b - B^{-1}Nx_N) + c_N^T x_N$$

$$= c_B^T B^{-1}b + \underbrace{(c_N^T - c_B^T B^{-1}N)}_{\lambda_N^T} x_N$$

$$= c_B^T B^{-1}b \quad (\because x_N = 0)$$

Suppose there exists a non-basic variable $x_q \in N$ ($x_q = 0$) such that $\lambda_q < 0$.

\therefore Objective Function at $x = c_B^T B^{-1}b + \lambda_q x_q$

The objective function can be decreased if x_q is changed from 0 to some positive value (by making x_q a basic variable).

MPTEL
Shrish Shekhar Numerical Optimization

Now, there is another way to look at this, and that depends that justification is based on the cost of the objective function, so if you can show that by making one of the non basic variable, a basic variable we can improve the objective function. So, that means, that there is a possibility of reduction in the objective function value, which means that the current basic feasible solution is not an optimal solution, so let us see how to do that. So, x is feasible implies that x equal to b and x nonnegative, so by making use of the components of A as B and N , we can write this as $B \times B$ plus $N \times N$ equal to b .

And therefore, B is x_B is equal to B inverse b minus B inverse $N \times N$, we have already seen this earlier, now let us look at the objective function of at x which is c transpose x . Now, as the variables are split in to two components basic and non basic, the objective function vector c is also split in to two components c_B and c_N , so transpose x

becomes $c B^T x_B$ plus $c N^T x_N$. And by using this value of x_B , we can write this as $c B^T B^{-1} b - B^{-1} N^T x_N$ plus $c N^T x_N$. And this is nothing but $c B^T B^{-1} b$ plus $c N^T - c B^T B^{-1} N^T x_N$, and this is nothing but the $\lambda^T x_N$ that we had seen earlier.

So, these are these are also called the relative cost factor corresponding to the non basic variables, and remember that the relative cost factor corresponding to the basic variables was 0, that $c B^T B^{-1} b$ that we saw earlier. And if $x_N = 0$ then current objective function is $c B^T B^{-1} b$, now this equation gives the clue about optimality. Now, if there exists some non basic variable say x_q , so that λ_q is less than 0, note that all the non basic variables are 0, so x_q also is 0, but if λ_q happens to be negative, then there will be a way to improve the objective function by making x_q to be positive.

So, let us see, so suppose there exists a non basic variable x_q in the set N such that, λ_q is less than 0, now since x_q is in the non basic variable set the current value of x_q is 0. Now, if you look at the objective function, objective function at x will be $c B^T B^{-1} b + \lambda_q x_q$, so let us assume that, we are not interested in the remaining non basic variables, their λ could be negative, positive or 0. But, we are at the moment interested only in one non basic variable which is x_q or which λ_q is less than 0.

So, the objective function can be written $c B^T B^{-1} b + \lambda_q x_q$, and remember that λ_q is less than 0, so currently the value of x_q is 0, now if we increase x_q to 0 to some positive quantity, the objective function value is going to decrease. Of course, we will move from x to some other new point in the process, but there is a scope for improvement in the objective function or decrease in the objective function, if x_q is increased from 0 to some positive quantity. And therefore, if x_q is changed from 0 to some positive quantity, we can say that x_q is made a basic variable.

Now, we have seen earlier that, a solution of a linear program lies at the boundary point and in a particular, if it lies at the boundary point, and if the solution set is convex and compact, then the solution lies at the extreme point also. So, from a current extreme point, if we want to move to an adjacent extreme point by making x_q as our basic variable, then one of the existing basic variables has to be eliminated or made non basic.

So, there will be a swap of a basic and a non basic variable between the sets B and N. So, that we can move from one vertex to an adjacent vertex, and in the process we will decrease the objective function value.

(Refer Slide Time: 15:37)

Suppose x_q is made a basic variable. Therefore, some existing basic variable $x_p (\in B)$ needs to be made *non-basic*.


$$Ax = b$$

$$\therefore Bx_B + Nx_N = b$$

$$\therefore x_B = B^{-1}b - B^{-1}Nx_N$$

$$\therefore \begin{pmatrix} x_1 \\ \vdots \\ x_p \\ \vdots \\ x_m \end{pmatrix} = \begin{pmatrix} \bar{b}_1 \\ \vdots \\ \bar{b}_p \\ \vdots \\ \bar{b}_m \end{pmatrix} - (B^{-1}N)_{\cdot,q} x_q$$

Note: $\bar{b} > \mathbf{0}$
 How to choose x_p ?
 Require $(B^{-1}N)_{p,q} > 0$ so that x_p can decrease if x_q increases.



Shrish Sengupta Numerical Optimization

Now, how to find out a variable which can or a basic variable which can become non basic, now let us assume that x_q is made a basic variable. And we are interested in finding out some existing basic variable x_B which belongs to the set B, that needs to be made non basic. Because, we know that any at every vertex for a given linear program, there are m variables corresponding to the basis set, and N minus m variables corresponding to the non basic set.

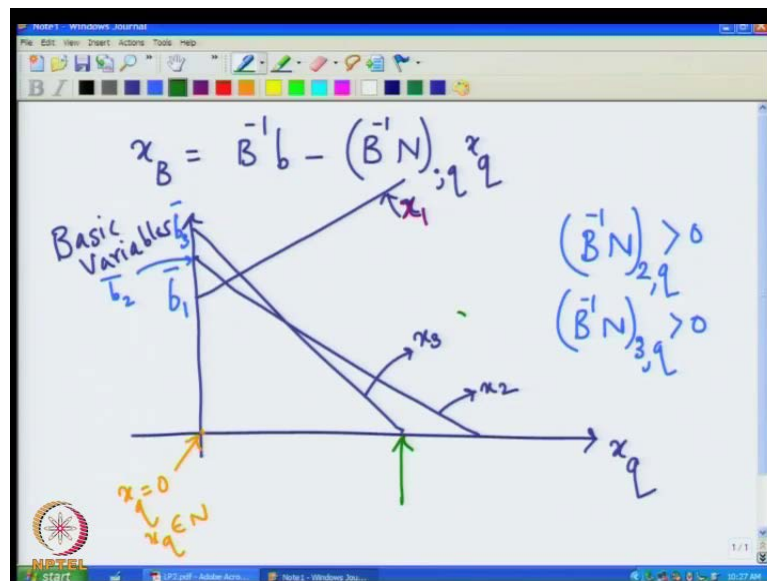
So, when we move to when we want to bring in a non basic variable in to the basis vector set, one of the existing basic variables has to be made non basic, and how that can be done that is the question that we would like to address now. So, again let us relook at the system $Ax = b$, and $Bx_B + Nx_N = b$, so we have $x_B = B^{-1}b - B^{-1}Nx_N$. Now, we are not interested in the entire $B^{-1}N$ matrix at this moment, because we are concentrating mainly on the one variable which is x_q , so x_q is currently non basic.

So, we are interested in the q th column of this $B^{-1}N$ matrix, so we can write x_B as $B^{-1}b - B^{-1}N \cdot$, I mean the q th column of $B^{-1}N$, so which is given here. So, you have x_1 to x_m as the basic variables without loss of

generality, now \bar{b}_1 to \bar{b}_m denote $B^{-1}b$, and since it is a non degenerate basic feasible solution, we can say that \bar{b} is greater than 0. And what we are interested in is the q th column of $B^{-1}N$, so that is denoted by $(B^{-1}N)_{i,q}$.

So, which means that we consider all the rows in the q th column of $B^{-1}N$ multiplied by x_q . So, at the moment we are interested only in the vary x_q variable of the non basic vector variable set, now let us look at this equation.

(Refer Slide Time: 18:36)



So, we have x_B is equal to $B^{-1}b$ minus $B^{-1}N$ (()), so this is the equation that we have, now so what we are looking at now is dependence of the basic variables x_B on the variables x_q which is currently 0. So, let us take x_q and on the vertical axis, we will consider all basic variables, now remember that currently x_q is equal to 0, because it is a non basic variable, so x_q belongs to N and it is non basic variable currently.

Now, there will be a variation in x_B , if x_q is a made a basic variable or in other words, if x_q is changed from the current value of 0 to some positive value, there will be a change in x_B . And that depends on the corresponding entry in the $B^{-1}N$ matrix q th column, so for example, one possibility is that for a particular variable, so let us call this variable as x_1 . So, so here we have \bar{b}_1 , \bar{b}_1 is nothing but the first entry in the vector $B^{-1}b$, and that corresponds to the first basic variable and without loss of generative erosion that the first m variables are basic variables.

So, what it means is that as x_q increases from 0 to some positive value, the basic variable x_1 also increases remember that, these are the basic variables x_B that we are plotting on the y axis, so as x_q increases x_1 also increases. Now, there could be another case where so for a variable 2, so this is corresponding to variable 2, so what it means is that as x_q increases the variable x_2 decreases, and comes to 0 at this value of x_q . Now, there could be a situation where, suppose that this corresponds to the variable x_3 , now x_3 behaves in the similar way to x_2 in the same sense that, as x_q increases the variable x_3 decreases.

So, this is b_3 bar and this value is b_2 bar, so you will see that as x_q is made a basic variable by increasing its value from 0 to some positive quantity, x_1 increases and both x_2 and x_3 decrease. Now, and both become 0 at certain value of x_2 , so for example, x_3 becomes 0 when x_q achieves this value, now among all these variables which are the candidate variables to be made non basic. Now, certainly x_1 is going x_q increases, so x_1 cannot be a candidate variable to be included to be included in the set N or to be excluded from the basic variable set.

Because, it is not going to become 0 as x_q increases on the other hand, x_2 and x_3 they become candidate variables as x_q becomes a basic variable, so they are the candidate variables to be made non basic. Now, out of x_1 , x_2 , x_3 you will see that these are the candidates where $b^{-1} B^{-1} N_q$ was greater than 0, so which means that the line, if you consider this as a line where x_2 denotes the horizontal axis, and the x_3 denotes the vertical axis, then that line has a negative slope. So, which means that $B^{-1} N$ has to be positive, so similarly $B^{-1} N$ that was also greater than 0.

So, the candidate basis vectors which possibly could be made non basic, is determined based on whether the corresponding entry in the $B^{-1} N$ matrix is positive or negative. So, for the variable x_1 the $B^{-1} N_{1,q}$ are the first entry in the q th column of $B^{-1} N$ matrix is negative, so which means that, the line x_B is equal to x_1 is equal to b_1 bar plus some quantity in to x_1 has a positive slope. On the other hand for these two variables x_2 and x_3 , this line has a negative slope and therefore, these are the candidates which can be made non basic, if x_q is made basic; now out of these candidates, there could be several more candidates like this, so out of all these candidates which one do we choose.

So, clearly we will see that, we will choose that basic variable which reaches the objective function which reaches the 0 value first, so clearly in this case where we have only three variables x_3 reaches 0 first compare to x_2 , because if x_2 increase further then x_3 becomes non positive or negative. And therefore, it will not remain feasible and therefore, we should choose this point where x_q or x_q is increased up to this point, so that x_3 becomes 0. And since, x_q is now positive, x_q can become a basic variable and x_3 which was in the previous iteration a basic variable, since it has attained a value of 0, it will become a non basic variable.

So, we are going to see this now, so the x_p is chosen based on the, q th the entries in the q th column of the B inverse N matrix, in particular x_p can be chosen such that, B inverse N p q has to be greater than 0, so that x_p can decrease if x_q increase. So, which means that if x_q increased from 0 to some positive value, x_p can be decreased from a current positive value to a value 0 which means that, x_p can be made non basic while making x_q a basic variable.

(Refer Slide Time: 28:23)

$$\begin{pmatrix} x_1 \\ \vdots \\ x_p \\ \vdots \\ x_m \end{pmatrix} = \begin{pmatrix} \bar{b}_1 \\ \vdots \\ \bar{b}_p \\ \vdots \\ \bar{b}_m \end{pmatrix} - (\mathbf{B}^{-1}\mathbf{N})_{:,q} x_q$$

How to choose the basic variable x_p to be made non-basic?
Let $\hat{B}_q = \{x_j \in B : (\mathbf{B}^{-1}\mathbf{N})_{jq} > 0\}$.

$$x_q = \min_{x_j \in \hat{B}_q} \frac{\bar{b}_j}{(\mathbf{B}^{-1}\mathbf{N})_{jq}} = \frac{\bar{b}_p}{(\mathbf{B}^{-1}\mathbf{N})_{pq}}$$

- If $\hat{B}_q = \phi$, the problem is *unbounded*.

NPTEL
Shrish Shevade Numerical Optimization

So, the natural question is that how to choose the basic variable, x_p to be made non basic. So, we saw earlier that we choose all the entries in they B inverse N matrix which are positive, and collect all those variables let us denote them by \hat{B}_q . So, \hat{B}_q denotes the set where denotes set of candidate basis vectors which could be made non basic. Now, among all those variables which are candidate basic vectors, we will try to

choose that basic vector which attains the value of 0 first or in other words, x_q will be assigned to the minimum $\bar{b}_p - (B^{-1}N)_{pq}$.

So, if you substitute x_p to be 0, we will get a value of x_q and among all those possible values of x_q will choose that value of x_q such that, the one of the basic variables becomes 0 first, that we saw in the illustration earlier. So, x_q will be $\bar{b}_p - (B^{-1}N)_{pq}$, so now, if you do this there are some possibilities that if you cannot get a non empty set \hat{B}_q , then the problem is unbounded. So, which means that we can increase the value of x_q without removing any of the basis vectors, and the objective function value goes to minus infinity or the problem is unbounded. So, if the problem is bounded, then x_q could be found using this formula or in other words if \hat{B}_q is non empty, then we can get a variable to be made non basic.

(Refer Slide Time: 30:51)

Summary of steps:
 Given a basic feasible solution, $(x_B \ x_N)^T$ w.r.t. B .
 If $\lambda_N = c_N - (B^{-1}N)^T c_B \leq \mathbf{0}$

- 1 Choose $x_q \in N$ such that $\lambda_q < 0$
- 2 $\hat{B}_q = \{x_j \in B : (B^{-1}N)_{jq} > 0\}$.
- 3 If $\hat{B}_q \neq \phi$, find the basic variable x_p to be made non-basic.

$$x_q = \min_{x_j \in \hat{B}_q} \frac{\bar{b}_j}{(B^{-1}N)_{jq}} = \frac{\bar{b}_p}{(B^{-1}N)_{pq}}$$

- 4 $x_i = \bar{b}_i - (B^{-1}N)_{iq}x_q, \forall x_i \in B$
- 5 Swap x_p and x_q from the sets B and N

Geometrical Interpretation of Steps 1-5
 Moving from one extreme point of the feasible set to an adjacent extreme point so that the objective function value decreases.

MPTEL Shrish Shevade Numerical Optimization

So, now, let us summarize the steps that we have seen earlier remember that, we started with the basic feasible solution which is $x_B \times N$ with respect to the basic matrix B . Now, if λ_N which is $c_N - (B^{-1}N)^T c_B$, if it less than or equal to 0 that means, the relative cost associated with the non basic variables are negative for at least one non basic variable. Then we choose one non basic variable x_q such that, λ_q is less than 0 remember that, when we say $\lambda \leq 0$, so which means that there exists at least one non basic variable whose relative cost is

negative, among all those variables whose relative cost are negative we choose one and denote it by λ_q .

Now, having chosen a variable or having chosen a non basic variable to be made basic the next step is to determine the variable or the basic variable to be made non basic. And for that purpose, we find out the matrix $B^{-1} \hat{q}$ sorry, the set $B^{-1} \hat{q}$, and that set is obtained by finding out all the basic variable for which $B^{-1} \hat{q}$ in the q th column is greater than 0. Or for all the basic variables x_j such that, the z_j entry in the q th column of B^{-1} matrix is greater than 0, and if that set is non empty, then the basic variable to be made non basic is obtained by finding out the minimum of b_j divided by $B^{-1} \hat{q}_j$, where x_j belongs to $B^{-1} \hat{q}$.

And that minimum value is given by the p th basic variable which will be made non basic, and which is a the value of 0 first and x_q is increased from 0 to some positive quantity. Now, once we do this what we need to do is that, we need to update all the basic variables and that is done by for every basic variable x_i , we update b_i minus $B^{-1} \hat{q}_i$ into x_q . So, if you this you will realize that when i is equal to p , we will be updating x_p to be b_p bar minus $B^{-1} \hat{q}_p$ into b_p bar minus $B^{-1} \hat{q}_p$, and which will make x_p to be 0.

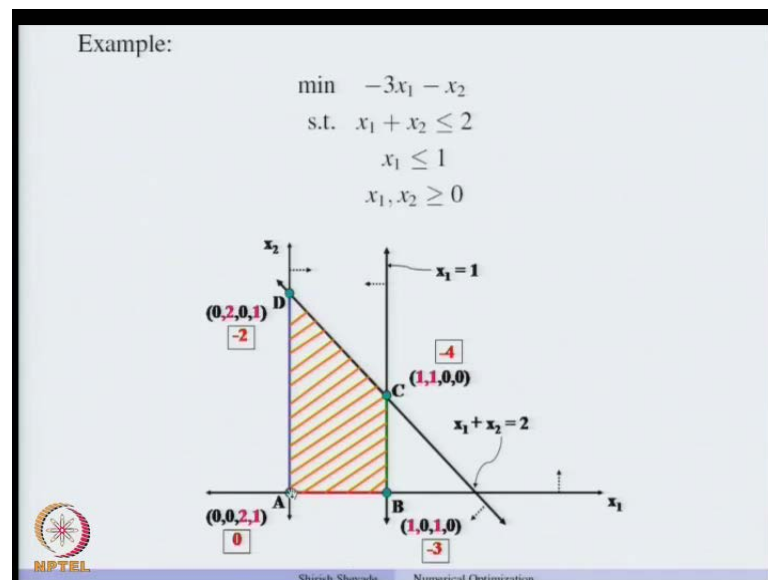
So, this update will automatically make x_p to be 0, this update will make x_q to be positive, and the remaining basic variables get updated using this, and remember that since we have chosen x_p to be made non basic. That means, that there will be a point where x_p becomes 0 and as x_p increases, and that is the first point where x_p becomes 0 and the rest of the non basic variables will remain 0. So, in the entire step what we did was, we increased x_q to a positive quantity, note that both the numerator and denominator are positive.

So, x_p increase, x_p to a positive quantity made it a basic variable by this update we made x_p 0, so that it becomes a non basic variable, and other basic variables they either increase or decrease depending upon the sign of $B^{-1} \hat{q}$. But, none of them will go to 0, the next step that we want to do is swap x_p and x_q from the sets B and N , just to indicate that now x_p has become a non basic variable, and x_q which was non basic earlier, has now become a basic variable. And this way of bringing in a basic a non basic variable leaving out one basic variable has a nice geometrical interpretation, and what it

denotes is that, is basically denotes how to move from one extreme point from the feasible set to an adjacent extreme point, so that the objective function value decreases.

So, remember that, we have chosen a variable x_q to be brought in to the basis vector set, based on the fact that $c^T B^{-1} b + \lambda q$ can be made negative, if x_q is made positive or it is a variable for which the relative cost factor λq is negative. So, so we have ensured that the objective function is going to decrease, if we make sure that x_q is increased, but then to move to the adjacent extreme points, one of the extreme points what we need to do is that, we need to make one of the basic variables are non basic variable; and that is done using this formula. So, this summary of steps has a nice geometric interpretation also.

(Refer Slide Time: 36:57)



And if you look at our example, which we saw in the last class, the problem is to minimize minus 3 x_1 minus x_2 , subject to x_1 plus x_2 less than or equal to 2, x_1 less than or equal to 1, and x_1 and x_2 non negative. So, you will see that this is a compact convex set, the feasible set is a compact convex set, the solution exists at one of the extreme points. Now, there are four extreme points or four vertices for this feasible set, and if you consider the point B suppose, then from B if we go to point A then the objective function value increases.

On the other hand, from B if we go to point C the objective function value decreases, and at C both the adjacent points are higher objective function value than the objective

function value at C which was which is minus 4. So, therefore, the solution to this problem is x_1 equal to 1 x_2 equal to 2, so these are the basic variables and this is also an optimal solution. Now, compare this optimal solution with the neighboring vertices objective function values, and also the basis vectors, so we will see that in moving from B to C the variable x_2 which was non basic here is made basic.

And variable x_3 which was basic is made non basic, similarly from this that the variable x_1 which was non basic is made basic while moving from D to C.

(Refer Slide Time: 39:25)

LP in Standard Form:

$$\begin{aligned} \min \quad & c^T x \\ \text{s.t.} \quad & Ax = b \\ & x \geq \mathbf{0} \end{aligned}$$

where $A \in \mathbb{R}^{m \times n}$ and $\text{rank}(A) = m$.

Given a basis matrix B w.r.t. the basis vector set B ,

- Basic Feasible Solution: $x_B = B^{-1}b$, $x_N = \mathbf{0}$
- Objective function = $c_B^T B^{-1}b$
- Relative cost factors, $\lambda_N = c_N - (B^{-1}N)^T c_B$

NPTEL
Shrish Shevade Numerical Optimization

Now, let us try to get an algorithm to solve this linear program in standard form, now let us assume that we are given a basis matrix B , which is associated with the basis vector set B . And the basic feasible solution x_B is equal to $B^{-1}b$ and x_N equal to 0, and assume that the basic feasible solution is nondegenerate, we assume that x_B is greater than 0. Now, if any of the B is negative, then by making some appropriate transformation, one can make sure that the initial basic feasible solution x_B is always greater than 0.

Now, we know that the objective function value is $c_B^T B^{-1}b$, because $x_N = 0$, so $c_N^T x_N$ is 0, so at a basic feasible solution $c_B^T B^{-1}b$ gives us an objective function value. Then the relative cost factors corresponding to the basic variables are 0, and the corresponding cost factors for the non basic variables are $c_N - B^{-1}N^T c_B$. Now, can we use this information and get away or

find a way, so that this the entire information is compactly available, in the form say matrix.

So, by looking at the matrix we should be able to find out what is B inverse b , what is or what is the current basic feasible solution, what is the objective function value which is $c^T B^{-1} b$, and what is the relative cost factors associated with the non basic variables. And remember that, the relative cost factors associated with the basic variables have to be 0, so if we have this information available, then that easily tells us whether we have reached an optimal solution.

If we have reached the optimal solution, then λ_N has to be greater than or equal to 0 for all non basic variables, and the if that is the case, then the current basic variables are x_B equal to $B^{-1} b$. So, this gives us the optimal solution, and the optimal cost is $c^T B^{-1} b$, so at a time if all this three pieces of information are available, then we can easily find out whether the given feasible solution is an optimal solution.

And this also would possibly help us to find out, if suppose λ_N is in not greater than or equal to 0 that means, there exists a non basic variable for which the Lagrangian multiplier is negative, or which has a negative cost factor. Then we should also be able to find out what is the basic variable that needs to be made non basic, so that the q th basic non basic variable can be made basic, so let us see how to do that.

(Refer Slide Time: 42:57)


LP in Standard Form:

$$\begin{aligned} \min \quad & c^T x \\ \text{s.t.} \quad & Ax = b \\ & x \geq 0 \end{aligned}$$

where $A \in \mathbb{R}^{m \times n}$ and $\text{rank}(A) = m$.

x_B	Basic Variables	Nonbasic Variables	RHS
	B	N	b
	c_B^T	c_N^T	0

$$\left(\begin{array}{c|c|c} I & B^{-1}N & B^{-1}b \\ \hline c_B^T & c_N^T & 0 \end{array} \right)$$

$$\left(\begin{array}{c|c|c} I & B^{-1}N & B^{-1}b \\ \hline 0^T & c_N^T - c_B^T B^{-1}N & -c_B^T B^{-1}b \end{array} \right)$$


Shreshth Sharma, NPTEL, Numerical Optimization

So, let us arrange the basic and non basic variables in the form of a matrix, so again without loss of generality we assume that the first m variables are basic variables, the next N minus m are non basic variables. And then we have the column to indicate the right hand of this, now first we write the matrix B associated with the basic variables, matrix N associated with the non basic variables; and then the corresponding vectors in the objective function. So, c transpose is related $c B$ transpose and $c N$ transpose, and this entry currently kept 0.

Now, you will see that this set of rows correspond to the basic variables, and what we want to do is that, we want to find out if the current feasible solution is a basic feasible solution that is x_B greater than 0 and x equal to 0. So, from this matrix we should be able to get what is the current basic feasible solution, and that is x_B equal to $B^{-1}b$ and we know that x_N equal to 0, so we do not have to worry about that part. So, how do we get $B^{-1}b$, so let us do some matrix operations, so multiply this first m rows or the matrix B by B^{-1} , so which will make the identity matrix.

Then this matrix will become $B^{-1}N$ matrix, and this will be this entry will be $B^{-1}b$, so so if the first m rows correspond to the that denote the basic variables, also refers to m columns denote the basic variables, so x_B is $B^{-1}b$. So, the current solution, current basic feasible solution is easy to obtain using this equal to $B^{-1}b$ and x_N equal to 0. Now, we also want the basic variables, cost factors, so that is $c N$ transpose minus B^{-1} minus $c B$ transpose $B^{-1}N$. So, what we do that, we multiply the m by N matrix I the m by N matrix here, by $c B$ transpose and subtract it from this.

So, what we get is $c B$ transpose minus $c B$ transpose I , which is nothing but $c B$ transpose minus $c B$ transpose and which is 0, and if pre-multiply $B^{-1}N$ by $c B$ transpose, and subtract it from $c N$ transpose. Then what we get is $c N$ transpose minus $c B$ transpose $B^{-1}N$, and here we get minus $c B$ transpose $B^{-1}b$. So, now, this is the information that we are looking for, so this matrix uses all the information that is needed to check whether a given feasible point is optimal, given basic feasible point is optimal.

So, for example, as we saw earlier that x_B is equal to $B^{-1}b$, so this the last column the last column denotes the value of x_B , and as I mentioned earlier $c N$ transpose minus

$c^T B^{-1} b$ is the nothing but λ^T . So, this allows us to check whether, the relative cost factors of non basic variables are nonnegative, now this 0 relates relative cost factor of the basic variables. Remember that, the first m columns here are associated with the basic variables, so the relative cost factors of basic variables are 0 , relative cost factors of non basic variables is $c^T N - \lambda^T N$.

And then in the last entry in this matrix is $-\lambda^T b$, now from this last entry we can easily find out what is $\lambda^T b$, which is basically the negative of the entry in this particular row and column; and that gives us the current objective function value. So, the basic feasible solution, the relative cost factors associated with non basic variables, and the objective function value, they all can be obtained by using this matrix.

(Refer Slide Time: 48:11)


Example:

$$\begin{aligned} \min \quad & -3x_1 - x_2 \\ \text{s.t.} \quad & x_1 + x_2 \leq 2 \\ & x_1 \leq 1 \\ & x_1, x_2 \geq 0 \end{aligned}$$

Given problem in the standard form:

$$\begin{aligned} \min \quad & -3x_1 - x_2 \\ \text{s.t.} \quad & x_1 + x_2 + x_3 = 2 \\ & x_1 + x_4 = 1 \\ & x_1, x_2, x_3, x_4 \geq 0 \end{aligned}$$

• Initial Basic Feasible Solution:
 $\mathbf{x}_B = (x_3, x_4)^T = (2, 1)^T, \mathbf{x}_N = (x_1, x_2)^T = (0, 0)^T$



Shreshth Shrivastava Numerical Optimization

So, let us see an example, again the same example that we saw to get a graphical solution to the problem. So, let us first write the linear program in standard form by introducing the slack variables for the two constraints, so now we have $x_1 + x_2 + x_3 = 2$ and $x_1 + x_4 = 1$. So, this is the linear program in standard form, because it of the type minimize $c^T x$ subject to $Ax = b, x \geq 0$. Now, one can easily determine it is initial basic feasible solutions, so what we need to is

that, you can just take x_3 to be 2 and x_4 equal to 1 or if you write in the form $Ax = b$, then we take a sub matrix of matrix A which is an identity matrix.

So, if you write the matrix A for this system of equations, so it will have two columns and four rows, so the first row is 1 1 1 0, and the second row is 1 0 0 1, so the last two columns of the matrix A they correspond to or the last 2 by 2 matrix of the matrix A , 2 by 2 sub matrix of A happens to be an identity matrix. So, we can easily get the initial basic feasible solution to be x_3 equal to 2 and x_4 equal to 1 by setting x_1 and x_2 to 0 or making x_1 and x_2 to be non basic.

(Refer Slide Time: 49:58)

$$\begin{aligned} \min \quad & -3x_1 - x_2 \\ \text{s.t.} \quad & x_1 + x_2 + x_3 = 2 \\ & x_1 + x_4 = 1 \\ & x_1, x_2, x_3, x_4 \geq 0 \end{aligned}$$

- Initial Basic Feasible Solution:
 $\mathbf{x}_B = (x_3, x_4)^T = (2, 1)^T$, $\mathbf{x}_N = (x_1, x_2)^T = (0, 0)^T$

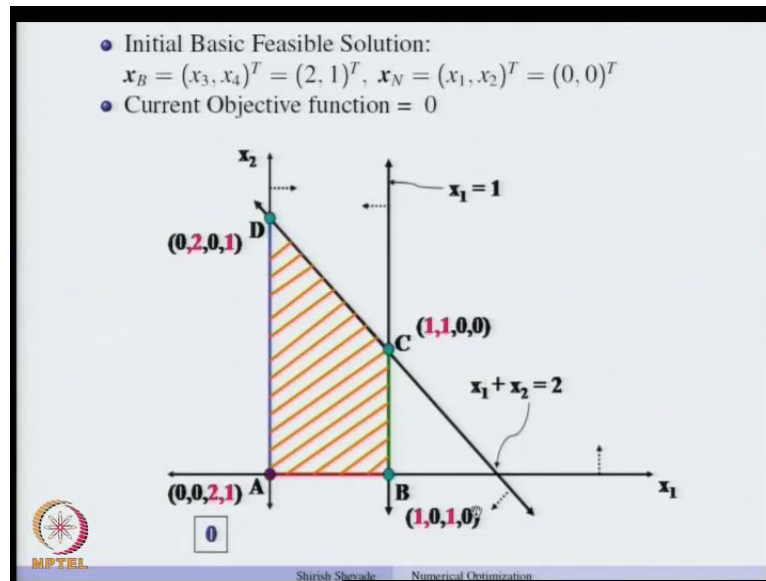
Initial Tableau:

x_1	x_2	x_3	x_4	RHS
1	1	1	0	2
1	0	0	1	1
-3	-1	0	0	0

MPTEL
 Shrish Shevade Numerical Optimization

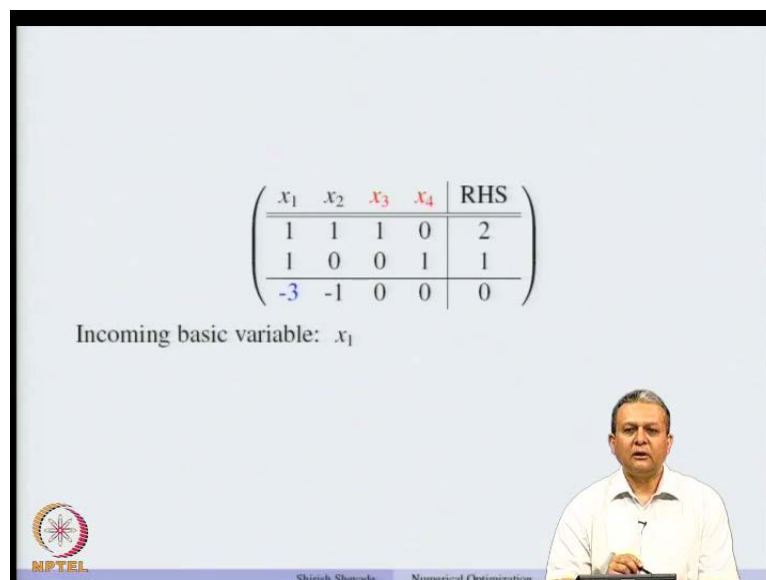
So, as I mentioned that this is the A matrix that we have got, associated with this constraint set, and the last two columns basically denote the identity matrix, so last two by 2 by 2 matrix of A , and that gives us the basic feasible solution. So, this row corresponds to x_3 , and this row corresponds to x_4 in this particular matrix, this is also called tableau. So, x_3 equal to 2 and x_4 equal to 1 is the current basic feasible solution, x_1 and x_2 are 0 and here, the relative cost factors of basic variables is 0; and the relative cost factors of the non basic variables are negative, so which means that the current feasible solution is not an optimal solution.

(Refer Slide Time: 51:24)



Now, to show it geometrically we have the point A, where x_1 and x_2 are 0 and x_3 and x_4 are 2, 1, so the basic variables are x_3 and x_4 , and the non basic variables are x_1 and x_2 , and the objective function value is 0 and that is what we saw from the tableau.

(Refer Slide Time: 51:55)



Now, the relative cost factors associated with the basic variables are negative, so we can choose any of these two variables to be made basic, now suppose we choose x_1 to be made basic; so that is indicated here by incoming basic variable x_1 , because it has a negative cost factor. Now, once we choose x_1 to be an incoming basic variable, we need

some basic variable to be made non basic. Now, currently x_3 and x_4 are basic variables, so if you look at the column associated with this minus 3, then you will see that all the entries are positive. So, this is the column in the matrix $B^{-1}N$ that is what we are looking at, so what we do is that, we determine the value \bar{b} by $B^{-1}N$. So, 2 by 1 and 1 by 1 , and among them we choose the one which has the lowest value this is also called ratio test.

(Refer Slide Time: 53:20)


$$\left(\begin{array}{cccc|c} x_1 & x_2 & x_3 & x_4 & \text{RHS} \\ \hline 1 & 1 & 1 & 0 & 2 \\ 1 & 0 & 0 & 1 & 1 \\ -3 & -1 & 0 & 0 & 0 \end{array} \right)$$

Incoming nonbasic variable: x_1

$$\left(\begin{array}{cccc|c} x_1 & x_2 & x_3 & x_4 & \text{RHS} \\ \hline 1 & 1 & 1 & 0 & 2 \\ 1 & 0 & 0 & 1 & 1 \\ -3 & -1 & 0 & 0 & 0 \end{array} \right)$$

Outgoing basic variable: x_4

$$\left(\begin{array}{cccc|c} x_1 & x_2 & x_3 & x_4 & \text{RHS} \\ \hline 0 & 1 & 1 & -1 & 1 \\ 1 & 0 & 0 & 1 & 1 \\ 0 & -1 & 0 & 3 & 3 \end{array} \right)$$

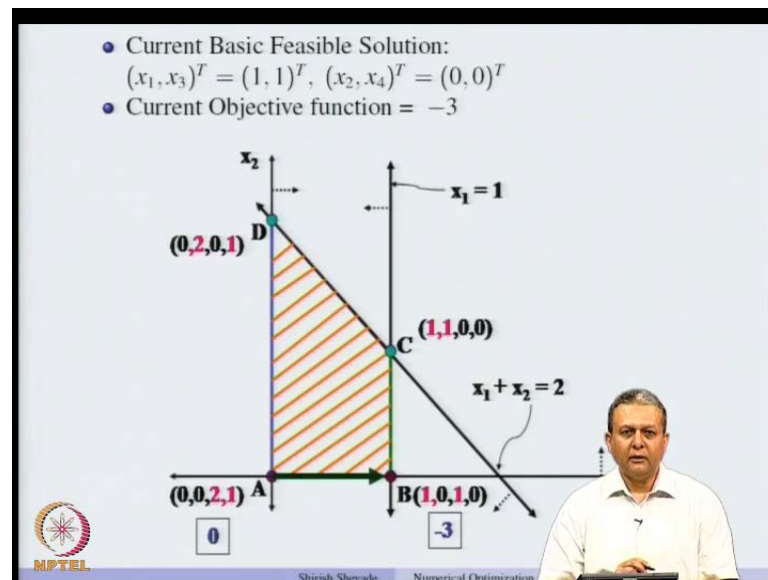

Shrish Shevade
Numerical Optimization

So, between the two ratios 2 by 1 and 1 by 1 , this ratio is the least which is shown by green, so the outgoing variable is corresponding to this row, and this row corresponds to the variable x_4 , so x_4 becomes an outgoing basic variable. So, we have incoming non basic variable which is x_1 , outgoing basic variable which is x_4 , now the next step is to make sure that, this column entire this column corresponds to the should be made similar to the column in the x_4 th variable case, so that means, that this entry should be made 0 , and this entry is 1 .

And then the relative cost of that variable, because it is going to become basic variable should be made 0 , so this is done using some matrix transformations, so for example, what we can do is that subtract the first row from the second row. So, we get 0 and then the remaining quantities, then subtract three times this row to the last row, so 3 into 1 plus minus 3 will make it 0 ; so which means that the basic variables cost, the relative cost if the basic variable is made 0 .

And now there will be corresponding changes in the column of x_4 , so now now if take if you look at the identity sub matrix, in this we will see that we will get x_1 and x_3 as the identity sub matrix of this. And which means that, x_1 and x_3 are basic variables, the value of x_3 is 1, value of x_1 is 1, and x_2 and x_4 are non basic. So, x_3 remains basic, x_4 which was basic earlier is now made non basic, and x_1 non basic earlier is made basic; and it was just ensured that the identity sub matrix of the this matrix is corresponding to the basic variables x_1 and x_3 . Now, this entry is 3, so which is which means that, the current objective function value is minus 3, so we have the basic feasible solution which is x_3 equal to 1 and x_1 equal to 1. The corresponding objective function value is minus 3, and the relative cost of the non basic variable x_2 is negative.

(Refer Slide Time: 56:28)



So, let us see this geometrically, so we have the current basic feasible solution which is (x_1, x_3) to be 1, 1, and (x_2, x_4) to be $(0, 0)$ current objective function value is minus 3, and this is indicated here. So, which means that by making x_1 to be basic, and removing x_4 from the basic variable set, we were able to move from the point A to the point B in this feasible region. And not only that, we also decrease the objective function value from 0 to minus 3 in moving that, so you will see that now x_1 and x_3 are the basic variables, and x_2 and x_4 are non basic. Now, the same procedure needs to be followed, and we will see that in the next class.

Thank you.