

Optimisation

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1 Whats reviewed

Lecture 1 ,2 and 3 tutorial 1,2,3 ,4, 5 ,6,7,8

2 Lecture 1_1

Notes: Linear function all components are linear(only x)
unbounded if a solution approaches infinity for maxi problem and negative infinity for minimization problem
linear programme- constraints and obj function linear

3 Tutorial 1

How a simple algebraic formulation might look like

· Let x_j be the number of ads purchased of type j for $j = 1$ to n .
Let a_j be the number of persons who view one ad of type i
total number of viewers must be at least b)
Let d_1 be an upper bound on the number of ads purchased of type j.
Minimize $\sum_{j=1}^n C_j x_j$
subject to $\sum_{j=1}^n a_{ij} x_j \geq b_i$
 $0 \leq x_j \leq d_j$ for $j = 1$ to n .

How a normal list of constraints might look like:

$$\begin{aligned}a_{11}x_1 + a_{12}x_2 &\leq b_1 \\a_{21}x_1 + a_{22}x_2 &\leq b_2 \\a_{31}x_1 + a_{32}x_2 &\leq b_3\end{aligned}$$

How it looks like in algebraic

$$\sum_{j=1}^n a_{ij} x_j \leq b_i \\ \text{for } i = 1 \text{ to } m.$$

n : the number of items.

a_{ij} : the amount of resource i used up by one unit of item j.

m : the number of different resources.

Standard form of L_p :

$$\begin{array}{ll} \text{maximize} & z = \sum_{j=1}^n c_j x_j \\ \text{subject to} & \sum_{j=1}^n a_{ij} x_j = b_i \quad \text{for } i = 1 \text{ to } m \\ & x_j \geq 0 \quad \text{for } j = 1 \text{ to } n \end{array}$$

4 Lecture 1_2

Least squares:

$$\min_{b_0, b_1, b_2 \in \mathbb{R}} \sum_{i=1}^6 (P_i - b_0 - b_1 L_i - b_2 E_i)^2 \quad \Bigg\}$$

$$\left. \begin{array}{l} \text{Linear Absolute Regression} \\ \forall i = 1, \dots, 6 \\ ri = 1, \dots, 6 \\ b_0, b_1, b_2 \end{array} \quad \begin{array}{l} \min \\ r_i^+ - r_i^- + b_0 + b_1 L_i + b_2 E_i \\ r_i^+, r_i^- \\ \in \mathbb{R} \end{array} \quad \begin{array}{l} \sum_{i=1}^6 (r_i^+ + r_i^-) \\ = \\ \geq 0 \end{array} \quad P_i \right\}$$

Maximum absolute residuals:

$$\left. \begin{array}{ll} \min & r \\ \forall i = 1, \dots, 6 & P_i - b_0 L_i - b_2 E_i \leq r \\ \forall i = 1, \dots, 6 & b_0 + b_1 L_i + b_2 E_i - P_i \leq r \\ r, b_0, b_1, b_2 & \in \mathbb{R} \end{array} \right\}$$

Converting to standard form

For free variables split into 2 variables plus variable 1 for the positive variable and minus negative variable 2

To convert from \geq or \leq to equal we add surplus/slack variables

- the variable if it is \geq

+ the variable if it is \leq

making sure all variables are ≥ 0 by introducing negative to flip the sign if necessary

RHS variables must be positive

adding artificial variables(more for to make into canonical)

1. $Ax^* = b$, and
2. there exists indices B_1, B_2, \dots, B_m such that:
 - (a) The m columns of matrix A , A_{B_1}, A_{B_2}, \dots , and A_{B_m} , are linearly independent;
 - (b) The decision variables not associated with the m linearly independent columns are zero, i.e., $x_i^* = 0$ for $i \neq B_1, B_2, \dots, B_m$

The set of variables $x_{B_1}, x_{B_2}, \dots, x_{B_m}$ is called a basis. The variables $x_{B_1}, x_{B_2}, \dots, x_{B_m}$ are called basic variables. The variables x_i for $i \neq B_1, B_2, \dots, B_m$ are called non-basic variables.

Consider an LP in the standard form

$$\left. \begin{array}{l} \max c^T x \\ \text{s.t.: } Ax = b \\ x \geq 0 \end{array} \right\}$$

where A is an $m \times n$ matrix with linearly independent rows and $b \in \mathbb{R}_+^m$. A vector x^* is a basic solution if

An LP is in canonical form if:

"It is in standard form (with possibly a constant term in the objective function).

Objective function, contains (a permutation of) the identity matrix as a submatrix.

The variables with a "+1" coefficient in the identity matrix appear with a zero coefficient in the objective function.

Hence we just add a row of artificial variables to make an identity matrix at the right.

Notes:

Lp standard form for this course is max

$m \times n$ matrix got n constraints and n linearly independent rows

linearly independent rows(basis)

x for non linearly independent row=0(non basic variable)

5 Tutorial 2

$$|x_1 + x_2 - x_3 - x_4| \leq 5$$

can be changed to

$$x_1 + x_2 - x_3 - x_4 \leq 5$$

and

$$-x_1 - x_2 + x_3 + x_4 \leq 5$$

Original equation:

$$\begin{array}{ll} \text{Minimize} & 500x_1 + 200x_2 + 250x_3 + 125x_4 \\ \text{subject to} & 50x_1 + 25x_2 + 20x_3 + 15x_4 \geq 1,500 \\ & 0 \leq x_1 \leq 20 \quad 0 \leq x_2 \leq 15 \quad 0 \leq x_3 \leq 25 \quad 0 \leq x_4 \leq 15 \end{array}$$

if we want to maximize the minimum of each constraint: we make the new variable $z \leq$ each of the components of the constraint and maximum that variable

$$\begin{aligned}
& \text{maximize } z \\
& \text{subject to } 50x_1 + 25x_2 + 20x_3 + 15x_4 \geq 1,500 \\
& \quad 0 \leq x_1 \leq 20 \quad 0 \leq x_2 \leq 15 \quad 0 \leq x_3 \leq 25 \quad 0 \leq x_4 \leq 15 \\
& \quad z \leq 50x_1, \quad z \leq 25x_2, \quad z \leq 20x_3, \quad z \leq 15x_4
\end{aligned}$$

If we want to minimize the maximum, we make the new variable \geq to each of the original obj functions and minimize that variable

$$\begin{aligned}
& \text{Minimize } Z \\
& \text{subject to } < \text{linear constraints} > \\
& \quad x \geq 0, y \geq 0 \\
& \quad z \geq 3x + 1 \\
& \quad z \geq 4y - 2
\end{aligned}$$

If we want to make a constraint where one variable must be a certain percentage of the whole amount produced: We can use this trick

$$\begin{aligned}
& x_1 / (x_1 + x_2 + x_3 + x_4) \geq 0.2 \\
& x_1 \geq 0.2 (x_1 + x_2 + x_3 + x_4) \\
& \text{Equivalently,} \\
& 0.8x_1 - 0.2x_2 - 0.2x_3 - 0.2x_4 \geq 0
\end{aligned}$$

But we need to make sure all variables in denominator are positive or 0, also make sure denominator does not add up to 0.

6 Tutorial 3

Basically a recap of conversion to standard form.

1. Standard Form must have non negativity constraints on all variables
2. All remaining constraints are equality constraints
3. RHS vector is non negative

Tips to convert

1. For non positive constraint on the RHS flip row of constraint (note: must be done before adding slack/surplus variable)
2. For \leq add +s
3. For \geq add -s
4. For non positive variables flip all variables
5. For free variables we can use 2 variables to sub that variable one +y1 -y2 (easy way)

6. Or use one constraint by making the variable the subject of the constraint to sub into all other constraints and obj function
7. to convert maximize to minimize and vice versa add negative sign on both sides

7 Tutorial 4

1. ERO 1: Multiply a row by a constant
2. ERO 2: Add a multiple of one row to another
3. ERO 3: Interchange two rows.

Try to perform gaussian elimination till you get an identity matrix or are unable to continue To pivot on entry (i, j) of a matrix is to carry out EROs so that

1. Row i is multiplied by a constant
2. Every other row has a multiple of Row i added to it.
3. After the operations, Column j has a 1 in Row i and a 0 elsewhere.

8 Tutorial 5

$$\begin{array}{ll} \text{maximize} & z = cx \\ \text{subject to} & Ax = b \\ & x \geq 0 \end{array}$$

This linear Program can be rewritten as

$$\begin{array}{ll} \max & \sum_{j=1}^n c_j x_j \\ \text{s.t} & \sum_{j=1}^n a_{ij} x_j = b_i \quad \forall i = 1 \text{ to } m \\ & x_j \geq 0 \quad \forall j = 1 \text{ to } n \end{array}$$

$-z$	x_1	x_2	x_3	x_4	x_5		RHS
1	c_1	c_2	c_3	c_4	c_5	=	$-z_0$
0	a_{11}	a_{12}	a_{13}	a_{14}	a_{15}	=	b_1
0	a_{21}	a_{22}	a_{23}	a_{24}	a_{25}	=	b_2
0	a_{31}	a_{32}	a_{33}	a_{34}	a_{35}	=	b_3

We put a bar over the symbols to indicate that it may not be the original tableau

$-z$	x_1	x_2	x_3	x_4	x_5		RHS
1	\bar{c}_1	\bar{c}_2	\bar{c}_3	\bar{c}_4	\bar{c}_5	=	$-\bar{z}_0$
0	\bar{a}_{11}	\bar{a}_{12}	\bar{a}_{13}	\bar{a}_{14}	\bar{a}_{15}	=	\bar{b}_1
0	\bar{a}_{21}	\bar{a}_{22}	\bar{a}_{23}	\bar{a}_{24}	\bar{a}_{25}	=	\bar{b}_2
0	\bar{a}_{31}	\bar{a}_{32}	\bar{a}_{33}	\bar{a}_{34}	\bar{a}_{35}	=	\bar{b}_3

We even have notation for the entering and leaving variables. The entering variable is typically denoted as variable x_s . For it to be an entering variable we need $\bar{c}_s > 0$.

Then we carry out the min ratio test to find the row on which to pivot. This row is typically given the index of r .

-2	x_1	x_2	x_3	x_4	x_5		RHS
1	\bar{c}_1	\bar{c}_2	\bar{c}_3	\bar{c}_4	\bar{c}_5	=	$-\bar{z}_0$
0	\bar{a}_{11}	\bar{a}_{12}	\bar{a}_{13}	\bar{a}_{14}	\bar{a}_{1s}	=	\bar{b}_1
0	\bar{a}_{r1}	\bar{a}_{r2}	\bar{a}_{r3}	\bar{a}_{r4}	\bar{a}_{rs}	=	\bar{b}_r
0	\bar{a}_{31}	\bar{a}_{32}	\bar{a}_{33}	\bar{a}_{34}	\bar{a}_{3s}	=	\bar{b}_3

9 lecture 2_1

Given an LP in a canonical form associated with a basis, the coefficient with which a variable appears in the objective function is called the reduced cost of that variable associated with that basis.

Simplex Algo:

1. Choose a positive reduced cost variable (Usually biggest one if we are greedy but may cycle)
2. Do ratio test ($\frac{\text{RHS coefficient}}{\text{Coefficient in constraint variable}}$) to find minimum this will be pivot
3. using this row convert the other rows to ensure that the coefficient of the (entering) variable in the rest of the row is 0, remember to change the entering basis variable coefficient to 1
4. Do this until all reduced cost are non positive (optimal) or until all the coefficient of a positive reduced cost is negative (unbounded)

At the end of phase 1,

1. if objective function is non zero then no feasible solution.
2. if obj function is zero
 - (a) no artificial variable in basis: we have BFS, drop artificial variable for BFS
 - (b) Some artificial variable in basis, pivot out, if cannot there are redundant constraints we can drop, then drop artificial variables and obtain BFS

This BFS forms the BFS for phase 2. change your objective function back to original and solve it using simplex algorithm again. We will get the answer as the obj function RHS and can read the value of variable from the RHS of the basis after we have finished the simplex algorithm.

Notes:

We increase objective function by increasing value of one of the non basic variable (we increase from 0 to some maximum value that another variable become 0 to replace it as the non basic variable), this causes the obj function to also increase as we like squeezing out some value out of the non basic variable and subbing another basic variable to sub it.

They are all related to each other by a ratio, so when one variable, the non basic variable increase, some other variables, the basic variable will become smaller, we do this till one of the non basic variable become 0 (not only those that are positive will become smaller, hence that's why we need at least one positive coefficient in a tableau to be like the limit, and if all variables are negative, we can increase to infinity thus unbounded), then we sub out that point for the rest of pts.

Intuition (at first is like got a value times $x=0$ cus they non basic variable hence like obj function is 0)

Later we move around such that is negative coeff for non basic variable and basic variable and positive coeff (at rhs) for basic coeff hence obj function maximized. Simplex algorithm is in essence just moving around the corners of a region.

Intuition why we do till all negative reduced cost for no basic variable because they times $x=0$ then 0 so don't decrease objective function.

10 Tutorial 6

A basic feasible solution is called degenerate if one of its RHS coefficients (excluding the objective value) is 0.

When a corner point is the solution of two different sets of equality constraints, then this is called degeneracy. This will turn out to be important for the simplex algorithm.

We say that a basis is degenerate if a basic variable is 0. (easiest definition)

Bland's Rule. (to ensure no cycling)

1. The entering variable should be the lowest index variable with negative reduced cost.
2. The leaving variable (in case of a tie in the min ratio test) should be the variable with the lowest index.

Finally, degeneracy is similar to but different from the condition for alternate optima. In degeneracy, one of the RHS values is 0. For alternate optima, in an optimal tableau one of the non-basic cost coefficients is 0.

11 Lecture 2_2

Eliminate artificial variables to get initial BFS

A simple tableau such that at least one of the right-hand side values is zero is called degenerate, and the corresponding BFS is degenerate.

Possible to change basis but BFS stays the same, pivot does not move point and may cause cycling

Bland Rule

1. Among variables with positive reduced cost, choose the one with the smallest index as the entering variable.
2. If there is a tie in the min ratio test, choose the one with the smallest index as the leaving variable.

$$\left. \begin{array}{ll} \max & c_B x_B + c_N x_N \\ A_B x_B + A_N x_N & = b \\ x_B, x_N & \geq 0 \end{array} \right\}$$

where c , c_b and c_N are row vectors

c_B refers to cost of basic variables

c_N refers to cost of non-basic variables

A_B is the basis matrix

We perform elementary row operations to obtain the RREF of the augmented matrix for the equality constraints. Equivalently, multiply A_B^{-1} on both sides of the equation.

$$\left. \begin{array}{ll} \max & c_B x_B + c_N x_N \\ I x_B + A_B^{-1} A_N x_N & = A_B^{-1} b \\ x_B, x_N & \geq 0 \end{array} \right\}$$

Finally, eliminate the basic variables in the objective function using $x_B = A_B^{-1} b - A_B^{-1} A_N x_N$

$$\left. \begin{array}{ll} \max & (c_N - c_B A_B^{-1} A_N) x_N + c_B A_B^{-1} b \\ I x_B + A_B^{-1} A_N x_N & = A_B^{-1} b \\ x_B, x_N & \geq 0 \end{array} \right\}$$

Basic	x_B	x_N		
$(-z)$	0	$c_N - c_B A_B^{-1} A_N$	=	$-c_B A_B^{-1} b$
x_B	I	$A_B^{-1} A_N$	=	$A_B^{-1} b$

With respect to a basis with matrix A_B , we define:

- The simplex multipliers: $y_B = c_B A_B^{-1}$
- The reduced costs of x_N : $\bar{c}_N = c_N - c_B A_B^{-1} A_N = c_N - y_B A_N$
- The reduced costs of x_B : $\bar{c}_B = 0$
- The objective function value $\bar{z} = c_B A_B^{-1} b = y_B b$
- $\bar{A}_N = A_B^{-1} A_N$
- The values of the basic variables: $\bar{b} = A_B^{-1} b$

Hence the tableau above can be defined as

Basic	x_B	x_N		
$(-z)$	0	\bar{C}_N	=	$-\bar{z}$
x_B	I	\bar{A}_N	=	\bar{b}

$$\left. \begin{array}{cccccccccccl} \max & 5x_1 & +2x_2 & +3x_3 & -4x_4 & +2x_5 & & & & & & \\ & & x_2 & & -x_4 & -x_5 & +x_6 & & & & & = 11 \\ & x_1 & & & & +x_5 & & +x_7 & & & & = 9 \\ -x_1 & -x_2 & +x_3 & +x_4 & & & & & x_8 & & & = 4 \\ x_1, & x_2, & x_3, & x_4, & x_5, & x_6, & x_7, & x_8 & \geq & 0 & \end{array} \right\}$$

$$c_B = [2 \ 5 \ 3] \quad C_N = [-4 \ 2 \ 0 \ 0 \ 0]$$

Basic	x_1	x_2	x_3	x_4	x_5	x_6	x_7	x_8	Rhs
$(-z)$	0	0	0	-2	-1	-5	-8	-3	-139
x_2		1		-1	-1	1			11
x_1	1				1		1		9
x_3			1			1	1	1	24

Consider an LP of the form $\max \{cx | Ax \leq b, x \geq 0\}$, where $b \geq 0$ and its canonical form $\max \{cx | Ax + Is = b, x \geq 0, s \geq 0\}$, where s is the vector of slack variables.

With respect to a basis B , the reduced cost of a slack variable is the negative of the simplex multiplier for the corresponding constraint. The column in the tableau corresponding to the slack variable s_j , \bar{A}_{sj} is the j^{th} column of A_B^{-1} .

This is just saying that the simplex multiplier y_B is the negative of the reduced cost of the slack variables, i.e. variables (that coincidentally form a diagonal at the end. (to make equation slack)

In this case is the $[5 \ 8 \ 3]$

The revised simplex method in three steps:

1. Find a nonbasic variable s such that $\bar{c}_s = c_s - y_B A_s > 0$

If none exists stop: the tableau is optimal.

2. Compute $\bar{A}_s = A_B^{-1} A_s$. Find

$$r = \arg \min_{i=1, \dots, m} \left\{ \frac{\bar{b}_i}{\bar{a}_{is}} : \bar{a}_{is} > 0 \right\}$$

if all $\bar{a}_{is} \leq 0$ stop: the problem is unbounded.

3. x_s enters the basis, x_r leaves. Update: $x_B, A_B^{-1}, y_B, \bar{b}$.

1. can be interpreted as finding a cost after a pivot that is positive

2. can be interpreted as finding using the min ratio test to find an argument which is positive and minimum

Notes:

A_N refers to array of (N)on basic variables constraints coeff. so the big N refers to non basic variables, small B refers to basic variables

12 Lecture 3_1

Do rmb those with a bar on top is the changed result after undergoing pivoting (for this lecture it refers to usually the final tableau cost), those without is the initial.

All questions asked a few slides back can be answered using the following formulas, that are fundamental for sensitivity analysis:

Simplex multipliers: $y_B = c_B A_B^{-1}$.

Optimal objective function value: $\bar{z} = y_B \bar{b}$.

Reduced cost: $\bar{c}_j = c_j - y_B A_j$.

Value of the basic variables in the BFS: $\bar{b} = A_B^{-1} b$.

Basic Tableau

Basic	x_1	x_2	x_3	x_4	x_5	x_6	Rhs
$(-z)$	52	30	20	0	0	0	0
x_4	10	5	2	1	0	0	204
x_5	2	4	5	0	1	0	100
x_6	1	1	1	0	0	1	30

Final Tableau

Basic	x_1	x_2	x_3	x_4	x_5	x_6	Rhs
$(-z)$	0	-2	0	-4	0	-12	-1176
x_1	1	0.375	0	0.125	0	-0.25	18
x_5	0	0.125	0	0.375	1	-5.75	4
x_3	0	0.625	1	-0.125	0	1.25	12

the set of basic variables: $B = \{x_1, x_5, x_3\}$

the basis matrix and its inverse:

$$A_B = \begin{bmatrix} 10 & 0 & 2 \\ 2 & 1 & 5 \\ 1 & 0 & 1 \end{bmatrix} \text{ and } A_B^{-1} = \begin{bmatrix} 0.125 & 0 & -0.25 \\ 0.375 & 1 & -5.75 \\ -0.125 & 0 & 1.25 \end{bmatrix}$$

the vector of simplex multipliers: $y_B = [4, 0, 12]$

the reduced costs: $\bar{c}_N = [-2, -4, -12]$ and $\bar{c}_B = [0, 0, 0]$

Here we have $b = (204, 100, 30)$ and $\Delta b = (\Delta b_1, 0, 0)$

$$x'_B = A_B^{-1}b + \Delta b_1[0.125, 0.375, -0.125]^T \geq 0$$

Therefore, we have a system of inequalities: $18 + \frac{1}{8}\Delta b_1 \geq 0$

$$4 + \frac{3}{8}\Delta b_1 \geq 0, \text{ and } 12 - \frac{1}{8}\Delta b_1 \geq 0, \text{ which yields}$$

$$-\frac{32}{3} \leq \Delta b_1 \leq 96$$

Expression for the reduced costs in tableau with basis B :

$$\bar{c}_N = c_N - y_B A_N$$

where $y_B = c_B A_B^{-1}$ are the shadow prices. As long as the basis and the cost of the basic variables c_B do not change, the shadow prices will stay the same.

When changing non basic variables, Since x_2 is nonbasic, if the basis remains optimal, change in c_2 affects only the reduced cost \bar{c}_2 .

$$\bar{c}'_2 = c_2 + \Delta c_2 - y_B A_2 = \Delta c_2 + \bar{c}_2 = \Delta c_2 - 2$$

note \bar{c}_2 can be read of final tableau

For basic variables, changing it will cost change in reduced cost for all non-basic variables.

We require the new reduced costs to be nonpositive, i.e.,

$$\begin{aligned} 0 &\geq \bar{c}'_j = c_j - (c_B + \Delta c_1 e_1) A_B^{-1} A_j \\ &= (c_j - c_B A_B^{-1} A_j) - \Delta c_1 e_1 A_B^{-1} A_j \\ &= \bar{c}_j - \Delta c_1 e_1 A_B^{-1} A_j \\ &= \bar{c}_j - \Delta c_1 e_1 \bar{A}_j \\ &= \bar{c}_j - \Delta c_1 \bar{a}_{1j} \end{aligned}$$

note \bar{a}_{1j} can be gotten from tableau

For $j = 2, 4, 6$, we have

$$\bar{c}'_2 = -2 - 0.375\Delta c_1 \leq 0$$

$$\bar{c}'_4 = -4 - 0.125\Delta c_1 \leq 0$$

$$\bar{c}'_6 = -12 - (-0.25)\Delta c_1 \leq 0$$

$$\text{This yields: } -\frac{16}{3} \leq \Delta c_1 \leq 48$$

(refer to tableau final above)

For new products (stool that require 2,1,1 processing and produce 12 in profit)

This is equivalent to adding one column to the problem:

$A_j = (2, 1, 1)^T$. The corresponding reduced cost is:

$$\bar{c}_j = c_j - y_B A_j, \text{ where } y_B = c_B A_B^{-1}$$

Because $y_B = (4, 0, 12)$, we obtain:

$$\bar{c}_j = 12 - (4, 0, 12)(2, 1, 1)^T = 12 - 8 - 12 = -8$$

It is not profitable to produce stools.

it is not profitable to produce because the reduced cost is $= -8$ which is not ≥ 0 , hence it will be a non basic variable in the function.

Notes:

if we change constraint RHS we just need make sure that each of the x is value is still positive after the change, this is due to the constraint where $x_1, x_2, x_3 \geq 0$, hence that's why value of x should not be negative.

obj function coefficient range for non basic variables is just change in cost \leq —reduced cost—

For basic variable need do some calculation to make sure its does not change bases.

hence use new reduced cost of non basic variable = old reduced cost + basic variable of interest row coeff for that non basic variable * change in obj coeff

13 Tutorial 7

The shadow price is also the derivative of the optimal value function $z(\Delta)$.

Determine the binding constraints, and determine the solution to the constraints if the RHS is increased by 1.

Suppose that we want to modify the cost coefficient for a variable x_j . We want to increase it from c_j to $c_j + \Delta$.

1. Determine the binding constraints and the current corner point solution, say x^* .
2. Compute the largest and smallest values of Δ so that the x^* remains optimal. In two dimensions, this will occur when the revised objective function is parallel to one of the constraints.

14 Tutorial 8

Using the 100% rule, we compute the amount of change divided by the total allowable change for each RHS that changes. In this case, divide the proposed decrease of 10 units of warehouse space by the 28 allowable decrease and do the same for juice glasses demand. Add up these fractions. If the total value is less

than 1, then the shadow prices are valid

There are a few different definitions for what reduced cost means. It is the shadow price for the non-negativity constraint. It is the objective value coefficient of a variable in the final and optimal tableau. And it is the objective coefficient obtained after pricing out the constraints.

15 Lecture 3_2

Unbounded problems have solutions where there is a positive reduced cost and non positive coefficients under that reduced cost.

Big M method

$$\left. \begin{array}{rcl} \max & x_1 + 3x_2 & \\ & 2x_1 - 2x_2 & \\ & x_1 + x_2 - s_1 & \\ & x_1, x_2, s_1, s_2, s_3 \geq 0 & \end{array} \right\}$$

Change the objective function to $x_1 + 3x_2 - Ms_2 - Ms_3$, where M is a very large number, and solve this new LP using the simplex method.

If there are alternative optimal solutions, This is noted by having a 0 reduced cost for a non basic variable in the final tableau, we can find 2 solutions of the equation and use a technique in order to get a line between them.

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \lambda \begin{pmatrix} 2 \\ 2 \\ 0 \end{pmatrix} + (1 - \lambda) \begin{pmatrix} 1 \\ 0 \\ 4 \end{pmatrix}$$

where the 2 matrix are the 2 solutions with equal objective function value

Notes:

To find alternative optimal solution use min ratio test on column of non basic variable where cost=0

Use convex combination to combine the vertex of all optimal solution.

16 Tutorial 9

For max problem

1. obj function coeff becomes constraint rhs values
2. max becomes min
3. constraint values become obj function coefficient
4. constraint coeff transposed
5. variables change

6. use SOB trick to change signs

check by using dual again, dual of dual equal primal. sensible odd bizzare

17 Disclaimer

Equations may have been copied wrongly from slides, Not liable for any exams. This can only be a last minute revision and secondary source in understanding the slides and classes. Any errors is purely my fault.(Please do not blame me too badly.)