TEC | Tecnológico de Costa Rica

Escuela de Ingeniería en Computación

Ingeniería en Computación

IC-6400 - Investigación de Operaciones

Project 05

Otro SIMPLEX más - Parte 2

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Chapter 1

Simplex Description

1.1 History

The Simplex algorithm is an algorithm discovered by George Dantzig. This algorithm is used find the optimal solution for a linear programming problem[1].Dantzig was an American mathmatician and Operations Research and Computer Science profesor at the Stanford University.He is called the father of Linear Programming thanks to the revolutionary Simplex algorithm.Before Dantzig it was known that the optimal solution to a linear programming problem could be found in a vertex of the feasible region,this made it posible to check every single vertex to find the solution but it was too slow to be used in practice.Then George Dantzig proved the feasible region is a convex shape, and after that he easily proposed a greedy algorithm that find the optimal solution to the problem.The property of being convex, allowed for the algorithm to be greedy and still be able to find the best solution.
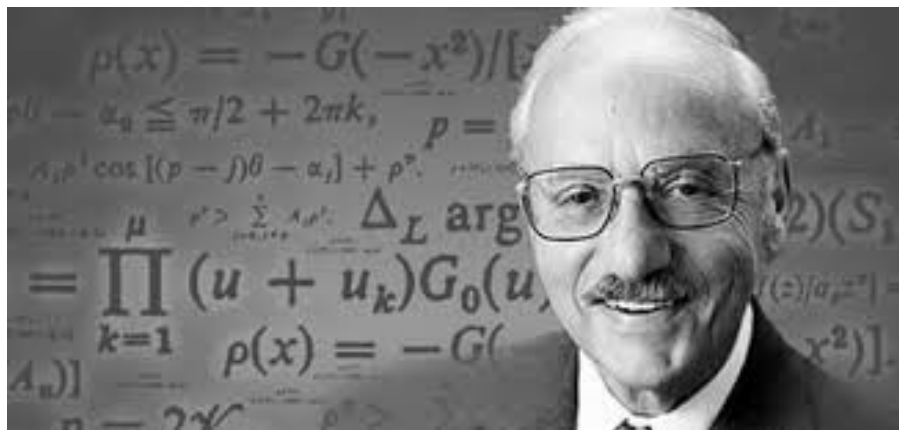

Figure 1.1: Image of George Dantzig

1.2 Description

The Simplex algorithm uses matrix operations to find the optimal solution to a linear programming problem, or to find out if it does not have a feasible solution [1].A linear programming problem has desition variables and restrictions under them, so you would think that you could solve the problem simply by using a normal method to find the value of the variables.Sadly, in these types of problems thre are usually far more varaibles than equations, to this aproach is not good enough.Using the equations you can find the feasible region of the problem, where all the posible solutions are.Simplex starts from a vertex of the feasible region and jumps to

adjacent vertices that give a better value for the objective function.Once you can not find a vertex that upgrades the objective value, that is the optimal solution.

The simplex algorithm has a set of variables that are called basic, these variables can be found in canonical columns [1].The first step is to level out any inecuations with slack or excess variables depending on what you need,when one of the restrictions is equals or greater than you will also need to add artificial variables, because these forms of restrictions do not give enough basic variables.After leveling out the restrictions, you need to clear out the z variable from the objective function, this will give you an extra equation to work with.

Once you have made that you can build the simplex table, giving each variable a column and each equation a row.To maximize the objective value you will find the smallest negative number in the first row (or the biggest positive number if you are minimizing).If you did not find any numbers that meet the criteria, congratulations, you found the optimal solution.However if you did, then you need to divide all the numbers in that column by the numbers of the final column (ignore negative numbers and 0's) and choose the smallest fraction to meet all the restrictions, the number on the column and row is called the pivot.Once you have a pivot you need to canonize the column, keep in mid that in this process you want the pivot to be the one of the canonical column, this process is called pivoting.Once you have canonized the column, repeat the process until you have ended [1].

Sadly, not every problem is as simplex(ba-dum-tss), if in the process of calculating the fractions you do not find any positive numbers, you have found an unbounded problem and it indicates that the objective value can keep growing to infinity, this usually indicates that you might be lacking a restriction.If at the end of a problem you see a non-basic variable that has a 0 in it's column it means that if you canonize that column with the apropiate pivot, the objective value will not go down, this indicates that the problem has multiple optimal solutions and you can choose any solution in between the initial two you found.If at any point of the execution of the algorithm you find a draw between which pivot to choose you have found yourself a degenerate problem, these are (in most cases) harmless, and you might see that the objective value does not upgrade in an iteration, do not worry, just keep pivoting.However, if you are really unlucky, you might find a degenerate problem that makes the algorithm enter an infinite loop, in these cases keep an eye on the tables, if they are repeating, you are indeed in a loop.But managing the extremely rare cases of the program entering a loop can be extremely expensive because you do not know when this will happen so you need to store every table of the execution, in these cases the usual solution is to look the other way and if a problem is taking too long to solve, just cut it's execution.

1.3   Big M Algorithm

The description of the simplex algorithm given in the last section seems simple enough, but it has a pretty big problem. The explanation given ealier only supports linear programming problems that have restrictions of less or equal, meaning that it cannot solve the ones that use greater or equal and equals. This problem arrives because of how you turn the inequations into equations, in less or equal restrictions you add some value as a slack variable, sice it sums to the restriction it is positive and thus produces a 1 in its position and a 0 in all other restrictions, making a canonical column. In greater or equal restrictions you use an excess variable, which subtracts from the restriction and thus produces a -1 on it's position, because of this it does not produces a canonical column Furthermore, equals restrictions do not have the need of extra variables, and thus they do not give canonical vectors. The problem here is that Simplex needs enough canonical vector to find the solution, but these types of restrictions do not produce canonical columns.

To deal with this you use the Big M Algorithm, this algorithm simply introduces the canonical

vectors needed, but it has a catch, when adding these, you need to add their value in the objective function as a M (-M if you are maximizind and +M if you are minimizing). Every restriction of the equals of greater of equal types, add a new artificial variable in order to have enough canonical vectors, when adding these to the table you will see that these vectors are not canonical, because they have M's in the first row, so you need to tackle this before anything. Once you build the simplex table, before execution simplex you need to canonize the columns of the artificial variables, you will see that other variables are starting to be filled with M's, but do not panic, this is completely normal. Once you have canonized all the columns of the artificial variables, you are free to use the Simplex algorithm, taking into accout that the M's should be interpreted as a very big value, a value so bigger than any other you have in the table. During the simplex execution you might see that the M's are starting to move out of the original variables and that they are staying in the artificial ones, which is good news. And take into account that once an artificial variable exits the base, it will never enter it again. So, what does the M means? As said before, the M is a value you can interpret as bigger than any other value in the table, and it adds dimensions to the problem to move around in them and hopefully find a solution that does not need the extra dimensions.

When using this algorithm you might see problems that never got the artificial variables out of their solution, or that still have M's but you cannot canonize it any further. When this happens, it means the solution is in another dimension, this means that the solution is out of the dimension of the original problem, and thus it means that the original problem has no feasible solution.

Chapter 2

Solving Multiple

2.1 Mathematical representation

Maximize:
$$z = 60,00000abeja + 35,00000babuino + 20,00000coyote \tag{2.1}$$

Subject To:

$$8,00000abeja + 6,00000babuino + 1,00000coyote \leq 48,00000 \tag{2.2}$$

$$4,00000abeja + 2,00000babuino + 1,50000coyote \leq 20,00000 \tag{2.3}$$

$$2,00000abeja + 1,50000babuino + 0,50000coyote \leq 8,00000 \tag{2.4}$$

$$0,00000abeja + 1,00000babuino + 0,00000coyote \leq 5,00000 \tag{2.5}$$

$$abeja, babuino, coyote \geq 0 \tag{2.6}$$

2.2 The initial simplex table

Z	$abeja$	$babuino$	$coyote$	s_1	s_2	s_3	s_4	B
1,00	-60,00	-35,00	-20,00	0,00	0,00	0,00	0,00	0,00
0,00	8,00	6,00	1,00	1,00	0,00	0,00	0,00	48,00
0,00	4,00	2,00	1,50	0,00	1,00	0,00	0,00	20,00
0,00	2,00	1,50	0,50	0,00	0,00	1,00	0,00	8,00
0,00	0,00	1,00	0,00	0,00	0,00	0,00	1,00	5,00

2.3 The intermediate simplex tables

Pivoting(1)

Z	$abeja$	$babuino$	$coyote$	s_1	s_2	s_3	s_4	B	$Fraction$
1,00	-60,00	-35,00	-20,00	0,00	0,00	0,00	0,00	0,00	–
0,00	8,00	6,00	1,00	1,00	0,00	0,00	0,00	48,00	6,00
0,00	4,00	2,00	1,50	0,00	1,00	0,00	0,00	20,00	5,00
0,00	2,00	1,50	0,50	0,00	0,00	1,00	0,00	8,00	4,00
0,00	0,00	1,00	0,00	0,00	0,00	0,00	1,00	5,00	–

Pivoting(2)

Z	$abeja$	$babuino$	$coyote$	s_1	s_2	s_3	s_4	B	$Fraction$
1,00	0,00	10,00	-5,00	0,00	0,00	30,00	0,00	240,00	–
0,00	0,00	0,00	-1,00	1,00	0,00	-4,00	0,00	16,00	–
0,00	0,00	-1,00	0,50	0,00	1,00	-2,00	0,00	4,00	8,00
0,00	1,00	0,75	0,25	0,00	0,00	0,50	0,00	4,00	16,00
0,00	0,00	1,00	0,00	0,00	0,00	0,00	1,00	5,00	–

2.4   The final simplex table

Z	$abeja$	$babuino$	$coyote$	s_1	s_2	s_3	s_4	B
1,00	0,00	0,00	0,00	0,00	10,00	10,00	0,00	280,00
0,00	0,00	-2,00	0,00	1,00	2,00	-8,00	0,00	24,00
0,00	0,00	-2,00	1,00	0,00	2,00	-4,00	0,00	8,00
0,00	1,00	1,25	0,00	0,00	-0,50	1,50	0,00	2,00
0,00	0,00	1,00	0,00	0,00	0,00	0,00	1,00	5,00

2.5   Multiple solutions found

Once you end the execution of the Simplex algorithm, you might think that everything has ended and that the soulution you found is the only way to get the optimal objective value, but this is untrue. Before asuming that you are finished check if you can pivot one of the non-basic variables without a penalty. To do this check all of the columns that correspond whith a non-basic variable, if one of them has a zero in the first row, this means that if you pivot it you objective value will not change. After you have found the row, canonize it, after doing it you will notice that as said, the objective value has not changed, but more importantly, you will find one of the basic variables is different, and that the values of the variables that stayed in the base have changed. This is another optimal solution to the problem, because the value is still optimal and the variables are different than before.
If you thought that this is the end, think again, remember that Simplex is a graphic method that jumps between vertices of the body that cointains all feasible solutions. Which means that the second solution you found is another vertex of the body. If you join the vertices that have an optimal solution, you will draw a line (or a plane, or another body in multiple dimensions). The body you drew contains other optimal solutions, infinite of them. This means that the problem you found can be optimized in an infinite number of ways.

By using the following formula, infinite optimal solutions can be found:

$$\alpha * solution1 + (1 - \alpha) * solution2 \tag{2.7}$$

$$0 \leq \alpha \leq 1$$

You can see the second optimal table found:

Z	$abeja$	$babuino$	$coyote$	s_1	s_2	s_3	s_4	B
1,00	0,00	0,00	0,00	0,00	10,00	10,00	0,00	280,00
0,00	1,60	0,00	0,00	1,00	1,20	-5,60	0,00	27,20
0,00	1,60	0,00	1,00	0,00	1,20	-1,60	0,00	11,20
0,00	0,80	1,00	0,00	0,00	-0,40	1,20	0,00	1,60
0,00	-0,80	0,00	0,00	0,00	0,40	-1,20	1,00	3,40

2.6   Solution

z=280,00
Solution 1:
$abeja = 2,00; babuino = 0,00; coyote = 8,00; s_1 = 24,00; s_2 = 0,00; s_3 = 0,00; s_4 = 5,00$
Solution 2:
$abeja = 0,00; babuino = 1,60; coyote = 11,20; s_1 = 27,20; s_2 = 0,00; s_3 = 0,00; s_4 = 3,40$
Solution 3:
$abeja = 0,50; babuino = 1,20; coyote = 10,40; s_1 = 26,40; s_2 = 0,00; s_3 = 0,00; s_4 = 3,80$
Solution 4:
$abeja = 1,00; babuino = 0,80; coyote = 9,60; s_1 = 25,60; s_2 = 0,00; s_3 = 0,00; s_4 = 4,20$
Solution 5:
$abeja = 1,50; babuino = 0,40; coyote = 8,80; s_1 = 24,80; s_2 = 0,00; s_3 = 0,00; s_4 = 4,60$

Bibliography

[1] F. Torres-Rojas, 'class about the simplex algorithm', personal communication, Investigación de Operaciones, Instituto Tecnológico de Costa Rica, San José, Costa Rica, Sep. 10, 2025.