# Duality Theorem A Tool For Quick Solution in Decision Making 

Adedoyin .I.S, Adeoye A. O., Olatinwo .M.<br>Department of Business Administration Kwara State Polytechnic, Ilorin, Nigeria<br>Department of Statistics Federal Polytechnic offa Kwara State Nigeria<br>Department of Mathematics Federal Polytechnic offa Kwara State Nigeria


#### Abstract

Linear programming problem is a mathematical technique for finding the best used of an organization resource, duality is a method of finding the solution to a linear programming problem. When considered linear programming problem with two variables using simplex method, the optimal solution is obtained at third iterations. When the problem is changed to dual problem and simplex method was used to obtained the optimal solution, the optimal solution is obtained at first iteration, and all information needed in the original problem are obtained in the dual result . This implied that dual problem has reduced the burden of computation by one iteration and all information needed in the original problem will be obtain at the optimal solution. This approach saves time and minimizes the cost incurred for estimating the maximum number of product expected to be produced by the management of industry.


Keywords: Duality theorem a better approach in linear programming

## I. INTRODUCTION

Duality In The Linear Programming Problem
Every Linear Programming Problem is associated with another Linear Programming Problem call the DUAL of the problem. The original problem is called PRIMAL, while the other is called DUAL. The optimal solution of either problem reveals' information concerning the optimal solution of the other. If the optimal solution of either primal or dual is known 'the optimal solution of the other is also available.
Rules of changing primal problem to Dual problem
i. Changing the variables from $X_{1}$ to any convenient variable say W.
ii. Transposing the coefficients matrix A
iii. Interchanging the role of constant terms coefficients of the objectives function
iv. Reverting the inequality and
v. Minimizing the objective function instead of

## maximizing it.

## Methodology

Given a linear programming problem (LPP)
Optimize $z=\underline{C^{T} x}$

$$
\begin{aligned}
& \text { s.t. } A \underline{x} * \underline{b} \\
& \underline{x} \geq 0
\end{aligned}
$$

To expand the above equation we have

$$
\begin{equation*}
\text { ie, OptimizeZ } Z_{p}=c_{1} x_{1}+c_{2} x_{2}+\ldots+c_{n} x_{n} \tag{1}
\end{equation*}
$$

## s.t.



Where * assume, $=, \leq, \geq$.
The original general problem is called Primal problem.
When the primal problem is change to dual problem, we have the following

## Optimize

$$
\begin{aligned}
z_{D}= & \underline{b^{T} w} \\
\text { s.t. } & A^{T} w \quad * \underline{c} \\
& \underline{w} \geq 0
\end{aligned}
$$

* assume, $=, \leq, \geq$.

The above can be expanded as follows.
Optimize
$z_{D}=b_{1} w_{1}+b_{2} w_{2}+\ldots+b_{n} w_{n}$

## s.t.

$$
\begin{aligned}
& a_{11} w_{1}+a_{12} w_{2}+a_{13} w_{3}+\ldots+a_{1 n} w_{n} \\
& a_{21} w_{1}+a_{22} w_{2}+a_{23} w_{3}+\ldots+a_{2 n} w_{n}
\end{aligned}
$$

$$
a_{\mathrm{m} 1} w_{1}+a_{\mathrm{m} 2} w_{2}+a_{\mathrm{m} 3} w_{3}+\ldots+a_{\mathrm{mu}} w_{n} * c_{\mathrm{mu}}
$$

$$
w_{j} \geq 0
$$

Where ${ }^{*}$ assume, $=, \leq, \geq$.
For a case of two variable we have optimize

$$
z_{p}=c_{1} x_{1}+c_{2} x_{2}
$$

## s. $t$.

$$
\begin{array}{cc}
a_{11} x_{1}+a_{12} x_{2} & * b_{1} \\
a_{21} x_{1}+a_{22} x_{2} & * b_{2} \\
& \cdot \\
a_{m 1} x_{1}+a_{m 2} x_{2} & * b_{m} \\
\underline{x} \geq 0
\end{array}
$$

Changing the above to dual problem we have optimize
$z_{p}=b_{1} w_{1}+b_{2} w_{2}+\ldots+b_{n} w_{n}$
s. $t$.

$$
\begin{gathered}
a_{11} w_{1}+a_{21} w_{2}+\ldots+a_{m 1} w_{m} \\
a_{12} w_{1}+a_{22} w_{2}+\ldots+c_{1} \\
\underline{w} \geq 0 .
\end{gathered}
$$

We will observe that we normally apply elementary row operation for the solution of linear programming problem. $z_{p}$ has m rows, while $z_{D}$ has two rows, which make our computations to be easy.

## Application

Given a linear programming problem with two variables
$\operatorname{Max} z_{\rho}=3 x_{1}-2 x_{2}$

$$
\text { s.t. } \begin{aligned}
x_{1} & \leq 4 \\
x_{1}+x_{2} & \leq 6 \\
x_{1}+x_{2} & \leq 5 \\
x_{2} & \leq 1 \\
x_{1}, x_{2} & \geq 0 .
\end{aligned}
$$

In other to solve the primal problem above, the slack variables need to be added. Adding slack variable for the primal problem becomes;
$\operatorname{Max} z_{p}=3 x_{1}-2 x_{2}$

$$
\begin{array}{r}
x_{1}+x_{3}=4 \\
x_{1}+x_{2}+x_{4}=6 \\
x_{1}+x_{2}+x_{5}=5 \\
-x_{2}-x_{6}=1
\end{array}
$$

$\underline{x} \geq 0$.
The first table of the above problem is;

| BV | CB | $x_{1}$ | $x_{2}$ | $x_{3}$ | $x_{4}$ | $x_{5}$ | $x_{6}$ | $x B_{i}$ |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
| $x_{3}$ | 0 | 1 | 0 | 1 | 0 | 0 | 0 | 4 |
| $x_{4}$ | 0 | 1 | 1 | 0 | 1 | 0 | 0 | 6 |
| $x_{5}$ | 0 | 1 | 1 | 0 | 0 | 1 | 0 | 5 |
| $x_{6}$ | 0 |  |  |  |  |  |  |  |
| $z_{j}-c_{j}$ | -3 | 2 | 0 | 0 | 0 | 0 | $z=0$ |  |

To go from the initial table to the next, we perform row operations, but in the $2^{\text {nd }}$ table, most zeros will vanish and most of the elements of the $2^{\text {nd }}$ table will be fraction. We observe again that the basic is about 4 by 4 matrix before we can get the optimal solution we would have many iterations. For the dual problem of the primal problem, linear programming problem becomes.

$$
\begin{gather*}
\min z_{D}=4 w_{1}+6 w_{2}+5 w_{3}+w_{4} \\
\text { s.t. } w_{1}+w_{2}+w_{3}+0 w_{4} \geq 3  \tag{i}\\
0 w_{1}+w_{2}+w_{3}+w_{4} \geq-2  \tag{ii}\\
w \geq 0 .
\end{gather*}
$$

Multiply the objective function by -1 and introducing the surplus variable, then the equation becomes

$$
\begin{gathered}
\operatorname{Max} \quad z_{D}^{\prime}=-4 w_{1}-6 w_{2}-5 w_{3}-w_{4} \\
\text { s.t. } \quad w_{1}+w_{2}+w_{3}+0 w_{4}-w_{5}=3 \\
0 w_{1}-w_{2}-w_{3}+w_{4}-w_{6}=1 \\
\underline{w} \geq 0 .
\end{gathered}
$$

Then the first simplex table becomes ;

| BV | CB | $w_{1}$ | $w_{2}$ | $w_{3}$ | $w_{4}$ | $w_{5}$ | $w_{6}$ |  | $w_{B}$ |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
| $w_{1}$ | -4 | 1 | 1 | 1 | 0 | -1 | 0 | 3 |  |
| $w_{4}$ | -1 | 0 | -1 | -1 | 1 | 0 | -1 | 2 |  |
|  |  |  |  |  |  |  |  |  |  |
| $z_{j}-c_{j}$ | 0 | 3 | 2 | 0 | 4 | 1 | . |  |  |

Since $\quad z_{j}-c_{j} \geq 0$, the solution is optimal at first table with

$$
\begin{aligned}
w_{1}=3, & w_{4}=2 \\
z_{D}^{\prime} & =-12+2=-10 \\
z_{D}^{\prime} & =10
\end{aligned}
$$

To obtain variable $\underline{x}$ from the dual table we look for the slack or surplus variable at the $z_{j}-c_{j}$ row of the optimal table above. The first slack or surplus variable correspond to the first variable ${ }^{x_{1}}$, while second one correspond to the variable $x_{2}$. Substituting from the objective function $z_{p}$ of primal problem

$$
x_{1}=4, \quad x_{2}=1
$$

$$
\begin{aligned}
& z_{p}=3 x_{1}-2 x_{2} \\
& z_{p}=3(4)-2(1)=10 .
\end{aligned}
$$

## II. CONCLUSION

The solution obtained using dual problem has reduced the burden of computation from one iteration to the next. This approach saves time and minimize the cost of incurred for estimating the maximum number of products expected to be produced by any company .

## REFERENCES

[1]. Tar P. , Stagel B. , and Maros .I. (2017) "Parallel search paths for the simplex algorithm,"Central Eur. J.
[2]. Oper. Res., vol. 25, no. 4, pp. 967_984, Dec. 2017.
[3]. Spampinato D. G., (2009)`Linear optimization with CUDA," Norwegian Univ.Sci. Technol., Trondheim, Norway, Fall Project Rep., Jan. 2009, pp. 29_65.
[4]. Gade-Nielsen N. F., Dammann B., and Jørgensen J. B.,(2014) "Interior point methods on GPU with
[5]. application to model predictive control," Tech. Univ.Denmark, Lyngby, Denmark, Tech. Rep., 2014.
[6]. Maggioni M.(2016), "'Sparse convex optimization on GPUs," Ph.D. dissertation, Univ. Illinois, Chicago, Chicago, IL, USA, 2016.
[7]. Jung J. H. and O'Leary D. P.,(2008) - Implementing an interior point method for linear programs on a
[8]. CPU-GPU system," Electron. Trans. Numer. Anal., vol. 28, nos. 174_189, p. 37, 2008.
[9]. Dantzig G. B., (2013) Alternate Algorithm for the Revised Simplex Method: Using a Product Form for the Inverse. Santa Monica, CA, USA: Rand.

