ECON 6100

Problem Set 1

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- 1. Problem of shipping books.
 - (a) In canonical form, defining x as the number of books shipped from Novato to San Francisco, y as the number of books shipped from Novato to Sacramento, z as the number of books shipped from Lodi to San Francisco, and w as the number of books shipped from Lodi to Sacramento, this problem is

$$v_P(b) = \max -(5x + 10y + 15z + 4w)$$

s.t. $-x - z \le -600$
 $-y - w \le -400$
 $x + y \le 700$
 $z + w \le 800$
 $x, y, z, w \ge 0$

In matrix form, we could represent this problem as

$$v_P(b) = \max c \cdot x$$

s.t. $Ax \le b$
 $x \ge 0$

where

$$A = \begin{bmatrix} -1 & 0 & -1 & 0 \\ 0 & -1 & 0 & -1 \\ 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 \end{bmatrix} \quad \text{and} \quad b = \begin{bmatrix} -600 \\ -400 \\ 700 \\ 800 \end{bmatrix} \quad \text{and} \quad c = \begin{bmatrix} -5 \\ -10 \\ -15 \\ -4 \end{bmatrix}^T$$

(b) Using the same variables as in part (a), and introducing the slack variable $\lambda \in \mathbb{R}^4$, the problem in standard form is

$$v_{P}(b) = \max -(5x + 10y + 15z + 4w) + 0 \cdot \lambda$$
 s.t.
$$-x - z + \lambda_{1} = -600$$
$$-y - w + \lambda_{2} = -400$$
$$x + y + \lambda_{3} = 700$$
$$z + w + \lambda_{4} = 800$$
$$x, y, z, w \ge 0$$
$$\lambda \ge 0$$

In matrix form, we could represent this problem as

$$v_P(b) = \max c \cdot x + 0 \cdot \lambda$$

s.t. $Ax + I_4\lambda = b$
 $x \ge 0$
 $\lambda \ge 0$

where, as above,

$$A = \begin{bmatrix} -1 & 0 & -1 & 0 \\ 0 & -1 & 0 & -1 \\ 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 \end{bmatrix} \quad \text{and} \quad b = \begin{bmatrix} -600 \\ -400 \\ 700 \\ 800 \end{bmatrix} \quad \text{and} \quad c = \begin{bmatrix} -5 \\ -10 \\ -15 \\ -4 \end{bmatrix}^T$$

and I_4 is the 4×4 identity matrix.

- 2. Starting from a linear program
 - (a) To write this problem in canonical form, we need to first deal with the free variable y. Define $y = y^+ y^-$, and we can say that $y^+, y^- \ge 0$. Our problem in canonical form is

$$v_P(b) = \max c_x x + c_y y^+ - c_y y^-$$
s.t. $x \le 1$

$$x, y^+, y^- \ge 0$$

With matrices, we have that the problem is

s.t.
$$az \le b$$
$$z \ge 0$$

where

$$z = \begin{bmatrix} x \\ y^+ \\ -y^- \end{bmatrix}^T \qquad c = \begin{bmatrix} c_x \\ c_y \\ c_y \end{bmatrix} \qquad A = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \qquad b = 1$$

(b) Using the same variable replacement as above, and introducing the slack variable $\lambda \in \mathbb{R}$, the problem in standard form is

$$v_P(b) = \max c_x x + c_y y^+ - c_y y^- + 0 \cdot \lambda$$

s.t. $x + \lambda = 1$
 $x, y^+, y^- \ge 0$
 $\lambda \ge 0$

With matrices, we have that the problem is

$$v_P(b) = \max c \cdot z + 0 \cdot \lambda$$

s.t. $Az + I^3\lambda = b$
 $z \ge 0$
 $\lambda \ge 0$

where, as above,

$$z = \begin{bmatrix} x \\ y^+ \\ -y^- \end{bmatrix}^T \qquad c = \begin{bmatrix} c_x \\ c_y \\ c_y \end{bmatrix} \qquad A = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \qquad b = 1$$

3. The primal problem is

$$v_P(b) = \max x_1 + 2x_2$$

s.t.
$$x_1 + x_2 \le 4$$

$$x_1 + 3x_2 \le b$$

(a) The primal problem represented in standard form, attained by introducing variables $x_1^+, x_1^-, x_2^+, x_2^-, x_2^+, x_2^+, x_2^-, x_2^$

$$v_P(b) = \max x_1^+ - x_1^- + 2x_2^+ - 2x_2^-$$
s.t.
$$x_1^+ - x_1^- + x_2^+ - x_2^- \le 4$$

$$x_1^+ - x_1^- + 3x_2^+ - 3x_2^- \le b$$

$$x_1^+, x_1^-, x_2^+, x_2^- \ge 0$$

We have that the relevant matrices are

$$c = \begin{bmatrix} 1 \\ -1 \\ 2 \\ -2 \end{bmatrix}^T \qquad ; \qquad b = \begin{bmatrix} 4 \\ b \end{bmatrix} \qquad ; \qquad A = \begin{bmatrix} 1 & -1 & 1 & -1 \\ 1 & -1 & 3 & -3 \end{bmatrix}$$

The dual problem is defined as

$$v_D(c) = \min y \cdot b$$

s.t. $yA \ge c$
 $y \ge 0$

Which becomes

$$v_D(c) = \min 4y_1 + by_2$$
s.t.
$$y_1 + y_2 \ge 1$$

$$-y_1 - y_2 \ge -1$$

$$y_1 + 3y_2 \ge 2$$

$$-y_1 - 3y_2 \ge -2$$

$$y_1, y_2 \ge 0$$

(b) The constraint sets for the problems are in Figure 1 (on the next page), with the primal constraint set shaded in Green and the dual constraint set in Purple. Note that the dual constraint set is a point – it is defined precisely by the two equations $y_1 + y_2 = 1$ and $y_1 + 3y_2 = 2$, the intersection of which is exactly the point (0.5, 0.5). The optimal points are also represented. Since the Duality Theorem holds, the same optimal value is attained in both problems – in the primal, the value of 2.5 is attained at the point (5.5, -1.5), and in the dual the value of 2.5 is attained at the point (0.5, 0.5).

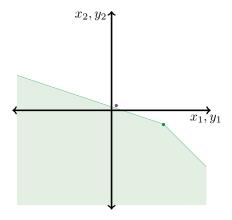


Figure 1: Constraint Sets

(c) The function $v_P(b)$ is, from the Vertex Theorem, always defined by the intersection of the functions $x_1 + x_2 = 4$ and $x_1 + 3x_2 = b$, where if $(x_1^{\star}, x_2^{\star})$ is a solution to that system, $v_P(b) = x_1^{\star} + 2x_2^{\star}$. Specifically, since we have that $x_2 = 4 - x_1$, we have that $x_1 = 6 - \frac{b}{2}$, and $x_2 = \frac{b}{2} - 2$, so

$$v_P(b) = 6 - \frac{b}{2} + b - 4 = 2 + \frac{b}{2}$$

Thus, for $b \in [0, 14]$, $\partial v_P(b) = \frac{1}{2}$. Both results are confirmed by solving the linear program for a discretization of that space, which leads to the plot in Figure 2.

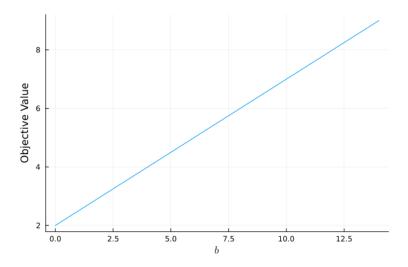


Figure 2: Objective value for a range of b, 0 to 14 with step size 0.01.

The code to create this figure, as well as to solve the linear program and the dual, is:

```
using JuMP, HiGHS, Plots, LaTeXStrings

function solve_lp(b)
    model = Model(HiGHS.Optimizer)

@variable(model, x[1:2])

@objective(model, Max, x[1] + 2x[2])

@constraint(model, x[1] + x[2] <= 4)
@constraint(model, x[1] + 3x[2] <= b)

set_silent(model)
    optimize!(model)

status = termination_status(model)
    if status == OPTIMAL
        return string(status), objective_value(model), value.(x)
    else
        return string(status), Float64[], Float64[]
    end</pre>
```

end

```
function solve_dual(b)
    model = Model(HiGHS.Optimizer)
   @variable(model, y[1:2] >= 0)
   @objective(model, Min, 4y[1] + b*y[2])
   (constraint(model, y[1] + y[2] == 1)
   (constraint(model, y[1] + 3y[2] == 2)
    set_silent(model)
    optimize!(model)
    status = termination_status(model)
   if status == OPTIMAL
        return string(status), objective_value(model), value.(y)
   else
        return string(status), Float64[], Float64[]
   end
end
b = 1
status, objective, solution = solve_lp(b)
println("Status: $status")
println("Objective value: $objective")
println("Solution: $solution")
dual_status, dual_objective, dual_solution = solve_dual(b)
println("Dual Status: $dual_status")
println("Dual Objective value: $dual_objective")
println("Dual Solution: $dual_solution")
range_of_b = 0:0.01:14
values = zeros(length(range_of_b))
for (i, b) in enumerate(range_of_b)
    values[i] = solve_lp(b)[2]
end
using Plots
plot(range_of_b, values, label="Objective Value", xlabel=L"b", ylabel=
   "Objective Value", legend=false)
savefig("ps1_objective_value.png")
```