# IDO- Tarea 3

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### Problema de la ONU

### **Enunciado**

ALGORITMO: Ramificación y Acotamiento 1. Inicializar la lista de nodos pendientes con el nodo raíz. 2. Mientras la lista de nodos pendientes no esté vacía, hacer: a. Seleccionar y eliminar un nodo de la lista de nodos pendientes. b. Si el nodo seleccionado es prometedor, entonces: i. Si el nodo seleccionado es una solución completa, comparar con la mejor solución actual. ii. De lo contrario, ramificar el nodo para generar nuevos nodos y agregarlos a la lista. 3. Devolver la mejor solución encontrada.

$$\begin{array}{ll} \max & z = 120x_1 + 80x_2 \\ & 2x_1 + x_2 + x_3 = 6 \\ & 7x_1 + 8x_2 + x_4 \leq 28 \\ & x_3 - .5x_4 = 2.5 \\ & x_i \in Z_+ \quad \text{para todo } i \end{array}$$

# Modelo en julia

Por método de ramificación y acotamiento se tiene el siguiente modelo en julia

Primero realizamos un simplex normal:

```
using JuMP, HiGHS
model = Model(HiGHS.Optimizer)
@variable(model, x[1:4] >= 0)

@objective(model, Max, 120x[1] + 80x[2])

@constraint(model, 2*x[1] + x[2] + x[3] == 6)
@constraint(model, 7*x[1] + 8*x[2] + x[4] <= 28)
@constraint(model, x[3] - 0.5*x[4] == 2.5)

optimization_result = optimize!(model)</pre>
```

```
Running HiGHS 1.6.0: Copyright (c) 2023 HiGHS under MIT licence terms Presolving model 2 rows, 3 cols, 6 nonzeros 2 rows, 3 cols, 6 nonzeros
```

```
Presolve: Reductions: rows 2(-1); columns 3(-1); elements 6(-2)
Solving the presolved LP
Using EKK dual simplex solver - serial
  Iteration
                   Objective
                                 Infeasibilities num(sum)
              -4.9998787442e+01 Ph1: 2(3.375); Du: 2(49.9988) Os
          0
               -2.8000000000e+02 Pr: 0(0) 0s
Solving the original LP from the solution after postsolve
Model
        status
                    : Optimal
Simplex
          iterations: 2
                   : 2.8000000000e+02
Objective value
HiGHS run time
                               0.00
El resultado es:
  value.(x)
4-element Vector{Float64}:
 -0.0
  3.5
  2.5
  0.0
```

#### Ramificación 1

Con lo cual decidimos ramificar en la variable  $x_3$ , dado que tuvimos soluciones fraccionadas:

#### Nodo 1

```
model = Model(HiGHS.Optimizer)
@variable(model, x[1:4] >= 0)

@objective(model, Max, 120x[1] + 80x[2])

@constraint(model, 2*x[1] + x[2] + x[3] == 6)
@constraint(model, 7*x[1] + 8*x[2] + x[4] <= 28)
@constraint(model, x[3] - 0.5*x[4] == 2.5)
@constraint(model, x[3] >= 3)

optimization_result = optimize!(model)
```

```
Running HiGHS 1.6.0: Copyright (c) 2023 HiGHS under MIT licence terms
Presolving model
2 rows, 3 cols, 6 nonzeros
2 rows, 3 cols, 6 nonzeros
Presolve: Reductions: rows 2(-2); columns 3(-1); elements 6(-3)
Solving the presolved LP
Using EKK dual simplex solver - serial
  Iteration
                   Objective
                                 Infeasibilities num(sum)
              -4.9998787442e+01 Ph1: 2(3.375); Du: 2(49.9988) Os
               -2.4000000000e+02 Pr: 0(0) 0s
Solving the original LP from the solution after postsolve
Model
        status
                   : Optimal
Simplex
          iterations: 3
Objective value : 2.4000000000e+02
HiGHS run time
                               0.00
  value.(x)
4-element Vector{Float64}:
 0.0
 3.0
 3.0
 1.0
Nodo 2
```

```
model = Model(HiGHS.Optimizer)
Ovariable(model, x[1:4] >= 0)
Objective(model, Max, 120x[1] + 80x[2])
0constraint(model, 2*x[1] + x[2] + x[3] == 6)
@constraint(model, 7*x[1] + 8*x[2] + x[4] <= 28)
Qconstraint(model, x[3] - 0.5*x[4] == 2.5)
@constraint(model, x[3] <= 2)</pre>
optimization_result = optimize!(model)
```

Running HiGHS 1.6.0: Copyright (c) 2023 HiGHS under MIT licence terms Presolving model

Problem status detected on presolve: Infeasible

Model status : Infeasible

ERROR: No LP invertible representation for getDualRay

#### Solución

Por lo tanto la solución es la del nodo 1, porque es la única que tiene solución entera de los dos espacios posibles, y sabiendo que en uno es infactible.

## Verdadera solución

De hecho la solución final del problema es:

```
using JuMP, HiGHS
  model = Model(HiGHS.Optimizer)
  @variable(model, x[1:4] >= 0, Int)
  Objective (model, Max, 120x[1] + 80x[2])
  @constraint(model, 2*x[1] + x[2] + x[3] == 6)
  0constraint(model, 7*x[1] + 8*x[2] + x[4] <= 28)
  @constraint(model, x[3] - 0.5*x[4] == 2.5)
  optimization_result = optimize!(model)
  value.(x)
Running HiGHS 1.6.0: Copyright (c) 2023 HiGHS under MIT licence terms
Presolving model
2 rows, 3 cols, 6 nonzeros
0 rows, 1 cols, 0 nonzeros
0 rows, 0 cols, 0 nonzeros
Presolve: Optimal
Solving report
  Status
                    Optimal
  Primal bound
                    240
  Dual bound
                    240
  Gap
                    0% (tolerance: 0.01%)
```

```
Solution status
                     {\tt feasible}
                     240 (objective)
                     0 (bound viol.)
                     0 (int. viol.)
                     0 (row viol.)
                     0.00 (total)
  Timing
                     0.00 (presolve)
                     0.00 (postsolve)
  Nodes
  LP iterations
                     0 (total)
                     0 (strong br.)
                     0 (separation)
                     0 (heuristics)
4-element Vector{Float64}:
0.0
 3.0
 3.0
```

1.0