

# Assignement 2

Sofia Davoli 813479

30/4/2020

## Problem 1

**1 Determine the nodes that will be visited by the BB algorithm and for each of them get the upper and lower limit deduced by the algorithm in the execution.**

B&B algorithm start from  $P_0$ , and split for  $x_1$ . if  $x_1=0$  then  $z_{best}=9$ , the algorithm consider then the other split relative to  $x_2, x_3$  but the solutions  $z_{best}=9$  have no improvement or is not feasible. So in the left part B&B stop at  $P_2$ .

if  $x_1=1$ , then  $z=16.2$  but the constraint are not satisfied, though other split are require for  $x_2$  and  $x_3$ . In the split of  $x_2$ , if  $x_2=0$ ,  $z_{best}$  have no improvement, if  $x_2=1$ ,  $z_{best}$  change to 16, the algorithm descend in this direction until the constraint of integrality are satisfied.

B&B stop in  $P_{11}$  where  $z_{best}=14$ ,  $x=(1,1,0,0)$  resulting in being the best solution for the problem.

Upper limit is obtain considering the integer part of the number in the previous split. (ex 17.8->17) Lower limit is obtain considering the possible minimum value the function can obtain in that precise split, with the previously assigned variables. (ex considering the objective function  $9x_1 + 5x_2 + 6x_3 + 4x_4$  in split  $P_{12}$  assigned variables are  $(x_1=1, x_2=0)$ , other variables can still vary from 1-0. The Lowest limit is equal to  $9 \times 1 + 5 \times 0 + 6 \times 0 + 4 \times 0 = 9$ )

limit for the split of the B&B algorithm are:

- $P_1$ : upper 16 lower 9

- $P_2$ : upper 16 lower 0

- $P_{12}$ : upper 16 lower 9

- $P_{14}$ : upper 13 lower 9

- $P_7$ : upper 16 lower 14

- $P_9$  upper 16 lower 14

**2 Solve the problem with an ILP solver and check the value of the objective function matches the one found at point 1.**

solving ILP problem with 0 constraint and 4 variables.

```
model = make.lp(0,4) # 0 constraints, 4 variables
lp.control(model, sense="max") # original maximization problem
```

```

set.objfn(model,obj=c(9,5,6,4)) # definition of the objective function using profit coef

# First constraint
add.constraint(model,
               xt=c(6,3,5,2),
               type="<=",rhs=10,
               indices=c(1,2,3,4))

# Second constraint
add.constraint(model,
               xt=c(1,1),
               type="<=",rhs=1,
               indices=c(3,4))

# Third constraint
add.constraint(model,
               xt=c(-1,1),
               type="<=",rhs=0,
               indices=c(1,3))

# Fourth constraint
add.constraint(model,
               xt=c(-1,1),
               type="<=",rhs=0,
               indices=c(2,4))

set.bounds(model, lower= c(0,0,0,0)) # non negativity
set.type(model, c(1:4), "binary")

```

Solution

```
solve(model)
```

```
## [1] 0
```

```
get.variables(model)
```

```
## [1] 1 1 0 0
```

```
get.objective(model)
```

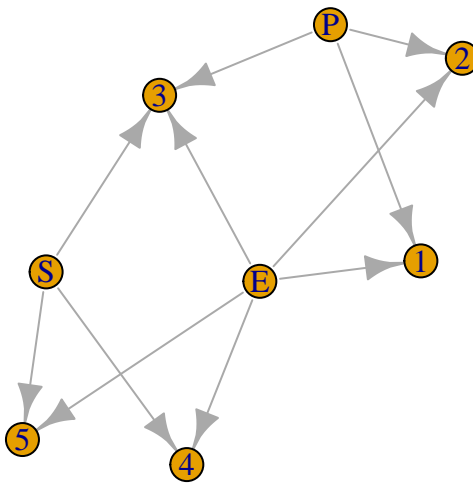
```
## [1] 14
```

The solution correspond to the one founded before.

## Problem 2

1 *Draw a network flow model to represent this problem.*

```
el <- matrix( c("P","1",6.50, "E","1",7.50, "P","2",7,
"E","2",8, "P","3", 8.25, "E","3", 7.25, "S","3", 6.75, "E","4", 7.75, "S","4", 7, "E","5", 7.50, "S",
nc = 3, byrow = TRUE)
g <- graph_from_edgelist(el = el[,1:2], directed = T)
edge_attr(g, "weight") <- el[,3]
plot(g, layout=layout_with_graphopt)
```



The length of the vector is proportional to it's cost.

2 *Implement your model and solve it.*

```
edges= data.frame(index_i=c("P", "E", "P","E", "P","E", "S", "E", "S", "E", "S"),
index_j =c(1,1,2,2,3,3,3,4,4,5,5),
lb =c(0,0,0,0,0,0,0,0,0,0,0),
ub =c(Inf,Inf,Inf,Inf,Inf,Inf,Inf,Inf,Inf,Inf,Inf),
cost =c(6.50, 7.5, 7, 8, 8.25, 7.25, 6.75, 7.75, 7, 7.50, 6.75)) #variables DF
b = c(30000, 40000, 25000, 35000, 33000) #customer per region
model = make.lp(0,11) #possibles linkes
```

```

lp.control(model, sense="min") #minimization problem
set.objfn(model,obj=edges$cost)

#constraint

set.bounds(model,lower=edges$lb,upper=edges$ub)

#link constraints
add.constraint(model,
  xt=c(1,1),
  type=">=",rhs=b[1],
  indices=c(1,2))

add.constraint(model,
  xt=c(1,1),
  type=">=",rhs=b[2],
  indices=c(3,4))

add.constraint(model,
  xt=c(1,1,1),
  type=">=",rhs=b[3],
  indices=c(5,6,7))

add.constraint(model,
  xt=c(1,1),
  type=">=",rhs=b[4],
  indices=c(8,9))

add.constraint(model,
  xt=c(1,1),
  type=">=",rhs=b[5],
  indices=c(10,11))

#capacity constraints
add.constraint(model,
  xt=c(1,1,1),
  type="<=",rhs=60000,
  indices=c(1,3,5))

add.constraint(model,
  xt=c(1,1,1,1,1),
  type="<=",rhs=70000,
  indices=c(2,4,6,8,10))

add.constraint(model,
  xt=c(1,1,1),
  type="<=",rhs=40000,
  indices=c(7,9,11))

set.type(model, c(1:11),type="integer") #model type

```

```
solve(model)
```

```
## [1] 0
```

```
get.variables(model)
```

```
## [1] 20000 10000 40000 0 0 25000 0 0 35000 28000 5000
```

```
get.objective(model)
```

```
## [1] 1155000
```

### 3 *What is the optimal solution?*

```
edges <- cbind(edges,y_val = get.variables(model))  
edges[c("index_i", "index_j", "y_val")]
```

```
##      index_i index_j y_val  
## 1          P      1 20000  
## 2          E      1 10000  
## 3          P      2 40000  
## 4          E      2      0  
## 5          P      3      0  
## 6          E      3 25000  
## 7          S      3      0  
## 8          E      4      0  
## 9          S      4 35000  
## 10         E      5 28000  
## 11         S      5  5000
```