Assignment 3

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Weiglet Primal Solution

```
Q1. Solve the problem using lpSolve or any other equivalent library in R.
```

```
#Weiglet Primal Solution
library(lpSolveAPI)
lprec < -make.lp(0,9)
lp.control(lprec, sense='max')
## $anti.degen
## [1] "fixedvars" "stalling"
## $basis.crash
## [1] "none"
## $bb.depthlimit
## [1] -50
## $bb.floorfirst
## [1] "automatic"
##
## $bb.rule
                                    "dynamic"
## [1] "pseudononint" "greedy"
                                                   "rcostfixing"
## $break.at.first
## [1] FALSE
## $break.at.value
## [1] 1e+30
## $epsilon
        epsb
                epsd
                          epsel
                                      epsint epsperturb epspivot
                             1e-12 1e-07 1e-05
##
       1e-10
               1e-09
                                                              2e-07
## $improve
## [1] "dualfeas" "thetagap"
## $infinite
## [1] 1e+30
## $maxpivot
## [1] 250
## $mip.gap
## absolute relative
     1e-11 1e-11
##
## $negrange
## [1] -1e+06
## $obj.in.basis
## [1] TRUE
##
## $pivoting
## [1] "devex"
                 "adaptive"
## $presolve
## [1] "none"
## $scalelimit
## [1] 5
## $scaling
## [1] "geometric" "equilibrate" "integers"
## $sense
## [1] "maximize"
## $simplextype
## [1] "dual" "primal"
## $timeout
## [1] 0
## $verbose
## [1] "neutral"
# Objective function with 9 decision variables
set.objfn(lprec,rep(c(420,360,300),3))
# Total there are 11 Constraints
```

```
add.constraint(lprec, c(1,1,1), "<=",750, indices = c(1,2,3))
 add.constraint(lprec, c(1,1,1), "<=", 900, indices = c(4,5,6))
 add.constraint(lprec, c(1,1,1), "<=", 450, indices = c(7,8,9))
 add.constraint(lprec, c(20, 15, 12), "<=", 13000, indices = c(1, 2, 3))
 add.constraint(lprec, c(20, 15, 12), "<=", 12000, indices = c(4, 5, 6))
 add.constraint(lprec, c(20, 15, 12), "<=", 5000, indices = c(7, 8, 9))
 add.constraint(lprec, c(1,1,1), "<=", 900, indices = c(1,4,7))
 add.constraint(lprec, c(1,1,1), "<=", 1200, indices = c(2,5,8))
 add.constraint(lprec, c(1,1,1), "<=",750, indices = c(3,6,9))
 add.constraint(lprec,c(rep(c(900, -750), each=3)), "=", 0, indices = c(1,2,3,4,5,6))
 add.constraint(lprec,c(rep(c(450,-750),each=3)),"=",0,indices = c(1,2,3,7,8,9))
Solve LP problem
 solve(lprec)
 ## [1] 0
Output of above solve is 0,So an optimal solution was obtained
```

Now we can get the value of the objective function, and the variables.

get.constraints(lprec)

get.variables(lprec)

[9] 416.6667

\$objfrom

##

price

lower

1 0.00 -1.000000e+30 1.000000e+30 ## 2 0.00 -1.000000e+30 1.000000e+30 ## 3 0.00 -1.000000e+30 1.000000e+30 ## 4 12.00 1.122222e+04 1.388889e+04 ## 5 20.00 1.150000e+04 1.250000e+04 ## 6 60.00 4.800000e+03 5.181818e+03 ## 7 0.00 -1.000000e+30 1.000000e+30 ## 8 0.00 -1.000000e+30 1.000000e+30 ## 9 0.00 -1.000000e+30 1.000000e+30 ## 10 -0.08 -2.500000e+04 2.500000e+04

Q4. Formulation of the dual of the above problem

Duality theory

lp.control(lprec, sense='min')

[1] "fixedvars" "stalling"

#Duality theory library(lpSolveAPI) lprec < -make.lp(0, 11)

\$anti.degen

[1] -50

\$break.at.first

\$break.at.value ## [1] -1e+30

[1] FALSE

##

[8] -1.00e+30 2.04e+02

Reduced cost mentioned in \$objfrom and \$objtill.

get.objective(lprec) ## [1] 696000

[1] 6.944444e+02 8.333333e+02 4.166667e+02 1.300000e+04 1.200000e+04 ## [6] 5.000000e+03 5.166667e+02 8.444444e+02 5.833333e+02 -2.037268e-10 ## [11] 0.00000e+00

[1] 516.6667 177.7778 0.0000 0.0000 666.6667 166.6667 0.0000

[1] 3.60e+02 3.45e+02 -1.00e+30 -1.00e+30 3.45e+02 2.52e+02 -1.00e+30

[11] -1.250000e+04 -1.000000e+30 -1.000000e+30 -2.222222e+02 -1.000000e+02 ## [16] -1.000000e+30 -1.000000e+30 -2.000000e+01 -4.444444e+01 -1.000000e+30

Medium Products and 166.67 of Small products. Plant 3: 416.67 of Small Products

Q2. Identify the shadow prices, dual solution and reduced costs. # Reduced Cost get.sensitivity.obj(lprec)

From the above solution, we can infer the following: Plant 1:516.67 of Large Products and 177.78 of Medium Products. Plant 2:666.67 of

0.0000

\$objtill ## [1] 4.60e+02 4.20e+02 3.24e+02 4.60e+02 4.20e+02 3.24e+02 7.80e+02 4.80e+02 ## [9] 1.00e+30

Shadow prices get.sensitivity.rhs(lprec) ## \$duals ## [1] 0.00 0.00 0.00 12.00 20.00 60.00 0.00 0.00 0.00 ## [10] -0.08 0.56 0.00 0.00 -24.00 -40.00 0.00 0.00 -360.00 ## [19] -120.00 0.00 ## ## \$dualsfrom ## [1] -1.000000e+30 -1.000000e+30 -1.000000e+30 1.122222e+04 1.150000e+04 ## [6] 4.800000e+03 -1.000000e+30 -1.000000e+30 -1.000000e+30 -2.500000e+04

\$dualstill ## [1] 1.000000e+30 1.000000e+30 1.000000e+30 1.388889e+04 1.250000e+04 ## [6] 5.181818e+03 1.000000e+30 1.000000e+30 1.000000e+30 2.500000e+04 ## [11] 1.250000e+04 1.000000e+30 1.000000e+30 1.111111e+02 1.000000e+02 ## [16] 1.000000e+30 1.000000e+30 2.500000e+01 6.666667e+01 1.000000e+30 Shadow prices are mentioned in \$duals, and the ranges for shadow price calculations present under \$dualsfrom and \$dualstill. Q3. Further, identify the sensitivity of the above prices and costs. That is, specify the range of shadow prices and reduced cost within which the optimal solution will not change s1 <- data.frame(get.sensitivity.rhs(lprec)\$duals[1:11],get.sensitivity.rhs(lprec)\$dualsfrom[1:11],get.sensitivity</pre> y.rhs(lprec)\$dualstill[1:11]) names(s1)<-c("price", "lower", "upper")</pre> s2 <- data.frame(get.sensitivity.rhs(lprec)\$duals[12:20],get.sensitivity.rhs(lprec)\$dualsfrom[12:20],get.sensitiv ity.rhs(lprec)\$dualstill[12:20]) names(s2)<-c("cost", "lower", "upper")</pre> s1

s2 ## cost lower upper ## 1 0 -1.000000e+30 1.000000e+30 0 -1.000000e+30 1.000000e+30 ## 2 ## 3 -24 -2.22222e+02 1.111111e+02 ## 4 -40 -1.000000e+02 1.000000e+02 0 -1.000000e+30 1.000000e+30 ## 6 0 -1.000000e+30 1.000000e+30 ## 7 -360 -2.000000e+01 2.500000e+01 ## 8 -120 -4.44444e+01 6.666667e+01 0 -1.000000e+30 1.000000e+30 ## 9 sensitivity of the above prices and costs are mentioned the range of shadow prices comprising of \$dualsfrom and \$dualstill.

\$basis.crash ## [1] "none" ## \$bb.depthlimit

The above is the range of shadow prices within which the optimal solution will not change.

```
## $bb.floorfirst
## [1] "automatic"
## $bb.rule
                                     "dynamic"
## [1] "pseudononint" "greedy"
                                                     "rcostfixing"
```

```
## $epsilon
                                          epsint epsperturb epspivot
         epsb
                  epsd
                               epsel
         1e-10
                  1e-09
                                1e-12
                                           1e-07
                                                      1e-05
                                                                   2e-07
 ## $improve
 ## [1] "dualfeas" "thetagap"
 ## $infinite
 ## [1] 1e+30
 ## $maxpivot
 ## [1] 250
 ## $mip.gap
 ## absolute relative
       1e-11
                1e-11
 ## $negrange
 ## [1] -1e+06
 ## $obj.in.basis
 ## [1] TRUE
 ## $pivoting
 ## [1] "devex"
                   "adaptive"
 ## $presolve
 ## [1] "none"
 ## $scalelimit
 ## [1] 5
 ## $scaling
 ## [1] "geometric" "equilibrate" "integers"
 ## $sense
 ## [1] "minimize"
 ## $simplextype
 ## [1] "dual" "primal"
 ## $timeout
 ## [1] 0
 ## $verbose
 ## [1] "neutral"
 # Objective function
 set.objfn(lprec,c(750,900,450,13000,12000,5000,900,1200,750,0,0))
 add.constraint(lprec, c(1, 20, 1, 900, 450), ">=", 420, indices = c(1, 4, 9, 10, 11))
 add.constraint(lprec, c(1, 15, 1, 900, 450), ">=", 360, indices = c(1, 4, 8, 10, 11))
 add.constraint(lprec,c(1,12,1,900,450),">=",300,indices = c(1,4,7,10,11))
 add.constraint(lprec, c(1, 20, 1, -750), ">=", 420, indices = c(2, 5, 9, 10))
 add.constraint(lprec, c(1,15,1,-750), ">=",360,indices = c(2,5,8,10))
 add.constraint(lprec, c(1, 12, 1, -750), ">=", 300, indices = c(2, 5, 7, 10))
 add.constraint(lprec, c(1, 20, 1, -750), ">=", 420, indices = c(3, 6, 9, 11))
 add.constraint(lprec, c(1, 15, 1, -750), ">=", 360, indices = c(3, 6, 8, 11))
 add.constraint(lprec, c(1, 12, 1, -750), ">=", 300, indices = c(3, 6, 7, 11))
 set.bounds(lprec,lower = c(-Inf,-Inf),columns = 10:11)
Solve LP problem
 solve(lprec)
 ## [1] 0
```

get.objective(lprec) ## [1] 696000 get.constraints(lprec) ## [1] 420 360 324 460 360 300 780 480 300

[1] 0.00 0.00 0.00 12.00 20.00 60.00 0.00 0.00 0.00 -0.08 0.56

[7] 583.3333 844.4444 516.6667 -25000.0000 -12500.0000

[15] 0.00000 316.66667 355.55556 233.33333 0.00000 0.00000

The above is the range of shadow prices within which the optimal solution will not change.

[1] 3.600000e+02 3.450000e+02 -1.0000000e+30 -1.000000e+30 3.450000e+02 ## [6] 2.880000e+02 -1.000000e+30 -1.000000e+30 2.040000e+02 -1.0000000e+30 ## [11] -5.915411e+13 -1.000000e+30 -1.000000e+30 -1.000000e+30 -1.000000e+30 ## [16] -2.400000e+01 -1.500000e+01 -4.000000e+01 -1.0000000e+30 -1.000000e+30

[1] 4.60e+02 4.20e+02 1.00e+30 1.00e+30 3.75e+02 3.24e+02 1.00e+30 1.00e+30 ## [9] 1.00e+30 2.52e+02 6.00e+01 4.80e+02 1.00e+30 1.00e+30 1.00e+30 1.20e+01

[1] 1.000000e+30 1.000000e+30 1.000000e+30 1.388889e+04 1.250000e+04 ## [6] 5.345455e+03 1.0000000e+30 1.0000000e+30 1.0000000e+30 1.0000000e+30

Output of above solve is 0,So an optimal solution was obtained

get.variables(lprec)

##

##

##

\$dualsfrom

\$dualstill

\$objtill

[11] 2.375000e+04

Now we can get the value of the objective function, and the variables.

reduced costs get.sensitivity.obj(lprec) ## \$objfrom ## [1] 694.4444 833.3333 416.6667 11222.2222 11050.0000 4800.0000

Reduced cost mentioned in \$objfrom and \$objtill. The above is the range of Reduced cost within which the optimal solution will not change. # shadow prices get.sensitivity.rhs(lprec) ## \$duals ## [1] 516.66667 177.77778 0.00000 0.00000 666.66667 166.66667 0.00000 ## [8] 0.00000 416.66667 55.55556 66.66667 33.33333 0.00000

The Formulation of dual solution agrees with primal solution.

Shadow prices are mentioned in \$duals, and the ranges for shadow price calculations present under \$dualsfrom and \$dualstill.

[17] 1.50e+01 6.00e+01 1.00e+30 1.00e+30