Question 1

For the optimizing values of x1, the three solvers found values close to 3, with Clp finding 3.0, while ECOS and SCS found values marginally below 3.0.

For the optimizing values of x2, the three solvers found values close to 0, with Clp finding 0.0, while ECOS and SCS found values marginally above 0.0.

For the optimizing values of x3, the three solvers found values close to 3, with Clp finding 3.0, while ECOS and SCS found values marginally above 3.0.

For the maximum value of the objective function, the three solvers found values close to 48, with Clp finding 48.0. ECOS found a value marginally below 48.0 and SCS found a value marginally above 48.0.

```
In [21]: # Solving using Clp
         using JuMP
         using Clp
         m = Model(solver = ClpSolver())
         @defVar(m, 0 \le x[1:3] \le 3)
                                                              #Variables x1,
          x2, x3
          @addConstraint(m, 2*x[1] >= x[2] +x[3])
                                                              #Constraint
          @setObjective(m, Max, 5*x[1] - x[2] + 11*x[3])
                                                              #Objective fun
         ction to maximize
         status = solve(m)
         println(m)
         println(status)
         println()
         println("x1 = ", getValue(x[1]))
         println("x2 = ", getValue(x[2]))
         println("x3 = ", getValue(x[3]))
         println("objective function = ", getObjectiveValue(m))
         Max 5 x[1] - x[2] + 11 x[3]
         Subject to
          2 x[1] - x[2] - x[3] \ge 0
          0 \le x[i] \le 3 \ \forall \ i \in \{1,2,3\}
         Optimal
         x1 = 3.0
         x2 = 0.0
         x3 = 3.0
         objective function = 48.0
```

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In []: -

```
In [6]: # Solving using ECOS
                          using JuMP
                          using ECOS
                         m = Model(solver = ECOSSolver())
                          @defVar(m, 0 \le x[1:3] \le 3)
                                                                                                                                                                                  #Variables x1, x2, x3
                          @addConstraint(m, 2*x[1] >= x[2] +x[3])
                                                                                                                                                                                  #Constraint
                          @setObjective(m, Max, 5*x[1] - x[2] + 11*x[3])
                                                                                                                                                                                  #Objective function to maximize
                          status = solve(m)
                          println(m)
                          println(status)
                         println()
                         println("x1 = ", getValue(x[1]))
                         println("x2 = ", getValue(x[2]))
                          println("x3 = ", getValue(x[3]))
                         println("objective function = ", getObjectiveValue(m))
                        \text{Max 5 } x[1] - x[2] + 11 x[3]
                         Subject to
                           2 x[1] - x[2] - x[3] \ge 0
0 \le x[i] \le 3 \ \forall \ i \in \{1,2,3\}
                         Optimal
                         x1 = 2.999999998571697
                        x2 = 8.223270011736391e-9
                         x3 = 3.000000001977236
                         objective function = 47.999999986810174
                        ECOS 2.0.4 - (C) embotech GmbH, Zurich Switzerland, 2012-15. Web: www.embotech.c
                         om/ECOS
                         Ιt
                                              pcost
                                                                                   dcost
                                                                                                                    gap
                                                                                                                                      pres
                                                                                                                                                           dres
                                                                                                                                                                                    k/t
                                                                                                                                                                                                         mu
                                                                                                                                                                                                                               step
                                                                                                                                                                                                                                                    sigma
                               IR
                                             BT
                            0 -2.250e+01 -8.440e+01 +1e+02 2e-01
                                                                                                                                                                                1e+00
                                                                                                                                                           3e-01
                          1 1 - | - -
                           1 -4.615e+01 -5.603e+01 +2e+01 2e-02 6e-02 7e-01
                                                                                                                                                                                                      3e+00
                         0 0 0 0 0
                            2 -4.726e+01 -4.850e+01 +3e+00 3e-03 8e-03 2e-01 4e-01
                                                                                                                                                                                                                         0.9283 7e-02
                         0 0 0 | 0 0
                            3 - 4.799e + 01 - 4.803e + 01 + 8e - 02 1e - 04 2e - 04 7e - 03 1e - 02
                                                                                                                                                                                                                        0.9798 9e-03
                         1 0 0 | 0 0
                            4 -4.800e+01 -4.800e+01 +9e-04 1e-06 3e-06 8e-05 1e-04
                                                                                                                                                                                                                        0.9890 1e-04
                         1 0 0 | 0 0
                            5 \quad -4.800 \\ \text{e} + 01 \quad -4.800 \\ \text{e} + 01 \quad +9 \\ \text{e} - 06 \quad 1 \\ \text{e} - 08 \quad 3 \\ \text{e} - 08 \quad 9 \\ \text{e} - 07 \quad 1 \\ \text{e} - 06 \quad 0.9890 \quad 1 \\ \text{e} - 04 \quad 0.9890 \quad 1 \\ \text{e}
                         1 0 0 | 0 0
                            6 -4.800e+01 -4.800e+01 +1e-07 1e-10 3e-10 1e-08 1e-08 0.9890 1e-04
                         1 0 0 | 0 0
                         OPTIMAL (within feastol=3.3e-10, reltol=2.2e-09, abstol=1.0e-07).
                         Runtime: 0.000080 seconds.
```

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```
In [7]: # Solving using SCS
        using JuMP
        using SCS
        m = Model(solver = SCSSolver())
        @defVar(m, 0 \le x[1:3] \le 3)
                                                           #Variables x1, x2, x3
        @addConstraint(m, 2*x[1] >= x[2] +x[3])
                                                           #Constraint
        @setObjective(m, Max, 5*x[1] - x[2] + 11*x[3])
                                                           #Objective function to maximize
        status = solve(m)
        println(m)
        println(status)
        println()
        println("x1 = ", getValue(x[1]))
        println("x2 = ", getValue(x[2]))
        println("x3 = ", getValue(x[3]))
        println("objective function = ", getObjectiveValue(m))
```

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```
\text{Max 5 } x[1] - x[2] + 11 x[3]
Subject to
2 x[1] - x[2] - x[3] \ge 0
0 \le x[i] \le 3 \ \forall \ i \in \{1,2,3\}
Optimal
x1 = 2.999985652990818
x2 = 4.149724928776938e-6
x3 = 3.0000130627112176
objective function = 48.00006780505256
______
      SCS v1.1.8 - Splitting Conic Solver
      (c) Brendan O'Donoghue, Stanford University, 2012-2015
    ______
Lin-sys: sparse-direct, nnz in A = 9
eps = 1.00e-04, alpha = 1.80, max iters = 20000, normalize = 1, scale = 5.00
Variables n = 3, constraints m = 7
Cones: linear vars: 7
Setup time: 1.53e-04s
______
Iter | pri res | dua res | rel gap | pri obj | dua obj | kap/tau | time (s)
_____
   0 inf inf nan -inf nan inf 2.57e-05
  100 8.00e-05 1.91e-04 8.48e-06 -4.80e+01 -4.80e+01 2.69e-15 1.50e-04
  140 | 4.49e-06 2.70e-06 1.09e-07 -4.80e+01 -4.80e+01 0.00e+00 2.06e-04
Status: Solved
Timing: Solve time: 2.08e-04s
      Lin-sys: nnz in L factor: 19, avg solve time: 2.01e-07s
      Cones: avg projection time: 5.10e-08s
Error metrics:
dist(s, K) = 1.3565e-17, dist(y, K*) = 0.0000e+00, s'y/m = -9.2050e-18
|Ax + s - b|_2 / (1 + |b|_2) = 4.4865e-06
|A'y + c|_2 / (1 + |c|_2) = 2.7040e-06
|c'x + b'y| / (1 + |c'x| + |b'y|) = 1.0946e-07
    ______
c'x = -48.0001, -b'y = -48.0001
   ______
```

In []:

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Question 2

a)

By the following substitutions:

$$z2 = x2 - 1,$$

 $z3 = x3 - 1,$
 $z4 = x4 - 2$

the bounds on z2, z3, z4 becomes:

$$0 \le x^2$$
, $x^3 \le 6$, $0 \le x^4 \le 4$

Introduce slack variables:

$$x2 + x2' = 6$$
, $x3 + x3' = 6$, $x4 + x4' = 4$, $x2'$, $x3'$, $x4' \Rightarrow 0$. (1)

These equations are added as constraints

z1 is transformed to:

$$z1 = x1 - x1',$$
 $x1, x1' => 0$

The inequality constraints are transformed into equality constraints by adding slack variables, x5 and x6, on the first and second original constraints:

$$-z1 + 6*z2 - z3 + z4 - x5 = -3,$$
 $x5 => 0$
 $7z2 + z4 = 5$
 $z3 + z4 + x6 = 2,$ $x6 => 0$

With substitution to x-variables the constraints become:

$$-x1 + x1' + 6*x2 - x3 + x4 - x5 = 4$$

 $7x2 + x4 = 14$
 $x3 + x4 + x6 = 5$

Add constraints from (1):

$$x2 + x2' = 6$$

$$x3 + x3' = 6$$

$$x4 + x4' = 4$$

In conclusion:

Original LP

```
In [3]: using JuMP
         m = Model()
         #Variables z1, z2, z3, z4
         @defVar(m, z1)
         @defVar(m, -1 \le z2 \le 5)
         @defVar(m, -1 \le z3 \le 5)
         @defVar(m, -2 \le z4 \le 2)
         #Constraints
         @addConstraint(m, - z1 + 6*z2 -z3 + z4 >= -3)
         @addConstraint(m, 7*z2 + z4 == 5)
         @addConstraint(m, z3 + z4 \le 2)
         #Objective
         @setObjective(m, Max, 3*z1 - z2)
         #Display results
         status = solve(m)
         println(m)
         println(status)
         println()
         println("z1 = ", getValue(z1))
         println("z2 = ", getValue(z2) )
         println("z3 = ", getValue(z3) )
         println("z4 = ", getValue(z4) )
         println("objective = ", getObjectiveValue(m) )
        Max \ 3 \ z1 - z2
        Subject to
         -z1 + 6 z2 - z3 + z4 \ge -3
         7 z2 + z4 = 5
         z3 + z4 \leq 2
         z1 free
         -1 \le z2 \le 5
         -1 \le z3 \le 5
         -2 \le z4 \le 2
        Optimal
        z1 = 8.571428571428571
        z2 = 0.42857142857142855
        z3 = -1.0
        z4 = 2.0
        objective = 25.28571428571429
In [ ]: -
```

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Transformed LP

In []:

```
In [12]: #c, A, b vectors
         c = [-3; 3; 1; 0; 0; 0; 0; 0; 0; 0]
         A = [-1 \ 1 \ 6 \ 0 \ -1 \ 0 \ 1 \ 0 \ -1 \ 0;
             0 0 7 0 0 0 1 0 0 0;
             0 0 0 0 1 0 1 0 0 1;
             0 0 1 1 0 0 0 0 0 0;
             0 0 0 0 1 1 0 0 0 0;
              0 0 0 0 0 0 1 1 0 0]
         b = [4; 14; 5; 6; 6; 4]
         using JuMP
         m = Model()
         @defVar(m, x[1:10] >= 0) #Variables: [x1; x1'; x2; x2'; x3; x3
          '; x4; x4'; x5; x6]
         @addConstraint(m, A*x .== b) #Constraint
         @setObjective(m, Min, dot(c, x) -1)
         #Calculate and display result
         println(solve(m))
         println()
         #Display original z-variables by reversing transformations on x-
         variables
         println("z1 = ", getValue(x[1] - x[2]))
         println("z2 = ", getValue(x[3] - 1))
         println("z3 = ", getValue(x[5] - 1))
         println("z4 = ", getValue(x[7] - 2))
         println("objective = ", -(getObjectiveValue(m)) ) #negated to gi
         ve maximum value
         Optimal
         z1 = 8.571428571428571
         z2 = 0.4285714285714286
         z3 = -1.0
         z4 = 2.0
         objective = 25.28571428571429
```

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Question3

The cost-minimizing steel has the following composition:

Material	Quantity(tons
Iron alloy 1	400.0
Iron alloy 2	0.0
Iron alloy 3	39.7763019923104
Copper 1	0.0
Copper 2	2.761272282418734
Aluminium 1	57.462425725270855
Aluminium 2	0.0

```
In [10]: #----#
         #Chemical elements
         element = [:carbon, :copper, :manganese]
         #Minimum grade of chemical elements required
         minGrade = Dict(:carbon => 0.02, :copper => 0.004, :manganese => 0.012)
         #Maximum grade of chemical elements required
         maxGrade = Dict(:carbon => 0.03, :copper => 0.006, :manganese => 0.0165)
         #Raw materials
         mat = [:ironAlloy1, :ironAlloy2, :ironAlloy3, :copper1, :copper2, :aluminium1, :a
         luminium2]
         #Carbon grade of raw materials
         carbGrade = Dict(:ironAlloy1 => 0.025, :ironAlloy2 => 0.03, :ironAlloy3 => 0,
                   :copper1 => 0, :copper2 => 0, :aluminium1 => 0, :aluminium2 => 0)
         #Copper grade of raw materials
         copGrade = Dict(:ironAlloy1 => 0, :ironAlloy2 => 0, :ironAlloy3 => 0.003,
                   :copper1 => 0.9, :copper2 => 0.96, :aluminium1 => 0.004, :aluminium2 =>
          0.006)
         #Manganese grade of raw materials
         mangGrade = Dict(:ironAlloy1 => 0.013, :ironAlloy2 => 0.008, :ironAlloy3 => 0,
                   :copper1 => 0, :copper2 => 0.04, :aluminium1 => 0.012, :aluminium2 => 0
         )
         #Amount of raw materials available in tons
         matQuant = Dict(:ironAlloy1 => 400, :ironAlloy2 => 300, :ironAlloy3 => 600,
                   :copper1 => 500, :copper2 => 200, :aluminium1 => 300, :aluminium2 => 25
         0)
         #Cost per ton of raw materials in dollars
         matCost = Dict(:ironAlloy1 => 200, :ironAlloy2 => 250, :ironAlloy3 => 150,
                   :copper1 => 220, :copper2 => 240, :aluminium1 => 200, :aluminium2 => 16
         5)
         #Amount of steel ordered in tons
         orderedQuant = 500
         #----#
         using JuMP
         m = Model()
         #Variables
         @defVar(m, matUsed[mat] >= 0)
                                                                                 #Amount o
         f a material used in tons
         #Expressions
         @defExpr(prodQuant, sum{matUsed[i], i in mat})
                                                                                      #Tot
         al material amount used in production
         @defExpr(totalCost, sum{matUsed[i]*matCost[i], i in mat})
                                                                                      #Tot
         al cost
         @defExpr(prodCarbGrade, sum{matUsed[i]*carbGrade[i], i in mat} / orderedQuant)
         #Carbon grade of the steel produced
         @defExpr(prodCopGrade, sum{matUsed[i]*copGrade[i], i in mat} / orderedQuant)
```

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```
Materials used:
matUsed: 1 dimensions, 7 entries:
  [aluminium1] = 57.462425725270904
  [aluminium2] = 0.0
  [ copper1] = 0.0
  [ copper2] = 2.761272282418734
  [ironAlloy1] = 400.0
  [ironAlloy2] = 0.0
  [ironAlloy3] = 39.77630199231035
Total cost: 98121.63579168123 dollars
```

In []:

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```
In [61]: #----#
          #Manufacturing methods
         method = [:m1, :m2]
          #Germanium grades
          grade = [:a, :b, :c, :d, :def]
          #Cost per transistor of the two methods and refiring in dollars
          cost = Dict(:m1 => 50, :m2 => 70)
          refCost = 25
          #Grade yields of the two methods in percentage
         yield = Dict(
          :m1 => Dict(:a => 0.05, :b => 0.15, :c => 0.2, :d => 0.3, :def => 0.3),
          :m2 => Dict(:a => 0.15, :b => 0.20, :c => 0.25, :d => 0.2, :def => 0.2)
          #Grade yields of the refiring process
          yieldR = Dict(
          :a \Rightarrow Dict(:a \Rightarrow 0.0, :b \Rightarrow 0.0, :c \Rightarrow 0.0, :d \Rightarrow 0.0, :def \Rightarrow 0.0),
          :b => Dict(:a => 0.5, :b => 0.5, :c => 0.0, :d => 0.0, :def => 0.0),
          :c => Dict(:a => 0.3, :b => 0.3, :c => 0.4, :d => 0.0, :def => 0.0),
          :d => Dict(:a => 0.2, :b => 0.2, :c => 0.3, :d => 0.3, :def => 0.0),
          :def => Dict(:a => 0.1, :b => 0.2, :c => 0.15, :d => 0.25, :def => 0.3),
          #Per month furnace capacity
          capacity = 20000
          #Demand for each grade of transistor
          demand = Dict(:a => 1000, :b => 2000, :c => 3000, :d => 3000, :def => 0)
Out[61]: Dict{Symbol,Int64} with 5 entries:
           :c => 3000
           :a => 1000
           :b => 2000
           :d => 3000
           :def => 0
```

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```
In [60]: # ----#
          using JuMP
          m= Model()
          #Method output
          @defVar(m, q[method] >= 0)
          #Refiring input
          @defVar(m, r[grade] >= 0)
          #Method grade output
          @defExpr(mlOut[g in grade], q[:m1]*(yield[:m1])[g])
          @defExpr(m2Out[g in grade], q[:m2]*(yield[:m2])[g])
          #Refiring grade output
          @defExpr(rOut[g in grade], sum{r[h]*yieldR[h][g], h in grade} )
          #Total transistors produced of each grade
          \texttt{@defExpr(finalOut[g in grade], mlOut[g] + m2Out[g] + rOut[g] - r[g])}
          #Total cost
          @defExpr(totalCost, sum{q[m]*cost[m], m in method} + sum{r[g], g in grade}*refCos
          t)
          #Constraints
          #Input into refiring is no more than output of method 1, 2
          @addConstraint(m, rIn[g in grade], r[g] \le m1Out[g] + m2Out[g])
          #Furnace capacity constraint
          \texttt{@addConstraint(m, sum}\{q[m], m \ \textbf{in} \ method} \ + \ sum\{r[g], \ g \ \textbf{in} \ grade\} \ <= \ capacity)
          #Demand fulfillment
          @addConstraint(m, meetDem[g in grade], finalOut[g] >= demand[g])
          #Minimize cost
          @setObjective(m, Min, totalCost)
          solve(m)
          println("Minimum cost required: ", getValue(totalCost), " dollars")
```

Minimum cost required: 641725.352112676 dollars

In []:

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```
In [78]: #----#
         #Electricity demand (MWH). Index corresponds to hour
         demand = [43, 40, 36, 36, 35, 38, 41, 46, 49, 48, 47, 47,
                   48, 46, 45, 47, 50, 63, 75, 75, 72, 66, 57, 50]
         purchaseLimit = 75
                                    #Maximum power can be purchased
         #Cost schedule
         cost1 = 100
         cost2 = 400
         cutOff = 50
                                   #Cut off point between first and second cost level
         batCap = 30
                                   #Battery power capacity
         #Calculate daily cost without battery
         totalCostWOBat = 0
         for i = 1:24
             if(demand[i] <= cutOff)</pre>
                 totalCostWOBat += demand[i]*cost1
                 totalCostWOBat += cutOff*cost1 + (demand[i] - 50)*cost2
             end
         end
In [79]: #----#
         using JuMP
         m = Model()
         @defVar(m, qBuy1[1:24] >= 0)
                                               #Electricity bounght on the first cost leve
         (defVar(m, qBuy2[1:24] >= 0)
                                               #Electricity bounght on the second cost lev
         @defVar(m, batLevel[1:25] >= 0)
                                               #Battery level at each hour
         @defExpr(totalCost, sum{qBuy1[i]*cost1 + qBuy2[i]*cost2, i in 1:24}) #Total cost
         @addConstraint(m, batLevel[1] == 0)
                                                                              #Initial batt
         ery level
         @addConstraint(m, batLevelLimit[i in 1:24], batLevel[i] <= batCap) #Battery leve</pre>
         1 must not exceed its capacity
         @addConstraint(m, priceLevel[i in 1:24], qBuy1[i] <= cutOff)</pre>
                                                                              #Cut off for
         the first price level
         #Electricty bought must be within the limit
         @addConstraint(m, avoidBlackout[i in 1:24], qBuy1[i] + qBuy2[i] <= purchaseLimit)</pre>
         #Flow constraint
         @addConstraint(m, balance[i in 1:24], qBuy1[i] + qBuy2[i] + batLevel[i] == demand
```

```
Total cost without battery: 152400 dollars
Total cost with battery: 143400.0 dollars
Savings: 9000.0 dollars
```

@setObjective(m, Min, totalCost) #Minimize cost

[i] + batLevel[i + 1])

solve(m)

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println("Total cost without battery: ", totalCostWOBat, " dollars")
println("Total cost with battery: ", getValue(totalCost), " dollars")
println("Savings: ", totalCostWOBat - getValue(totalCost), " dollars")

```
In [80]: #Model with infinite battery capacity
         #----#
         #Electricity demand (MWH). Index corresponds to hour
         demand = [43, 40, 36, 36, 35, 38, 41, 46, 49, 48, 47, 47,
                   48, 46, 45, 47, 50, 63, 75, 75, 72, 66, 57, 50]
         purchaseLimit = 75
                                    #Maximum power can be purchased
         #Cost schedule
         cost1 = 100
         cost2 = 400
         cutOff = 50
                                   #Cut off point between first and second cost level
         batCap = 30
                                   #Battery power capacity
         #Calculate daily cost without battery
         totalCostWOBat = 0
         for i = 1:24
             if(demand[i] <= cutOff)</pre>
                 totalCostWOBat += demand[i]*cost1
                 totalCostWOBat += cutOff*cost1 + (demand[i] - 50)*cost2
             end
         end
```

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```
In [98]: #----#
         using JuMP
         m = Model()
         (defVar(m, qBuy1[1:24] >= 0)
                                               #Electricity bounght on the first cost leve
         \thetadefVar(m, qBuy2[1:24] >= 0)
                                               #Electricity bounght on the second cost lev
         @defVar(m, batLevel[1:25] >= 0)
                                               #Battery level at each hour
         @defExpr(totalCost, sum{qBuy1[i]*cost1 + qBuy2[i]*cost2, i in 1:24}) #Total cost
         @addConstraint(m, batLevel[1] == 0)
                                                                              #Initial batt
         ery level
         #@addConstraint(m, batLevelLimit[i in 1:24], batLevel[i] <= batCap) #Battery lev
         el must not exceed its capacity
         @addConstraint(m, priceLevel[i in 1:24], qBuy1[i] <= cutOff)</pre>
                                                                              #Cut off for
         the first price level
         #Electricty bought must be within the limit
         @addConstraint(m, avoidBlackout[i in 1:24], qBuy1[i] + qBuy2[i] <= purchaseLimit)</pre>
         #Flow constraint
         @addConstraint(m, balance[i in 1:24], qBuy1[i] + qBuy2[i] + batLevel[i] == demand
         [i] + batLevel[i + 1])
         @setObjective(m, Min, totalCost)
                                              #Minimize cost
         solve(m)
         println("Total cost without battery: ", totalCostWOBat, " dollars")
         println("Total cost with battery: ", getValue(totalCost), " dollars")
         println("Savings: ", totalCostWOBat - getValue(totalCost), " dollars")
         println("Battery levels: ", getValue(batLevel'))
         Total cost without battery: 152400 dollars
         Total cost with battery: 120000.0 dollars
         Savings: 32400.0 dollars
         Battery levels: [0.0 7.0 17.0 31.0 45.0 60.0 72.0 81.0 85.0 86.0 88.0 91.0 94.0
```

At most 108.0 MW of battery power is used

96.0 100.0 105.0 108.0 108.0 95.0 70.0 45.0 23.0 7.0 0.0 0.0]

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